

# Calsequestrin-Mediated Mechanism for Cellular Calcium Transient Alternans

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**ABSTRACT** Intracellular calcium transient alternans (CTA) has a recognized role in arrhythmogenesis, but its origin is not yet fully understood. Recent models of CTA are based on a steep relationship between calcium release from the sarcoplasmic reticulum (SR) and its calcium load before release. This mechanism alone, however, does not explain recent observations of CTA without diastolic SR calcium content alternations. In addition, nanoscopic imaging of calcium dynamics has revealed that the elementary calcium release units of the SR can become refractory independently of their local calcium content. Here we show using a new physiologically detailed mathematical model of calcium cycling that luminal gating of the calcium release channels (RyRs) mediated by the luminal buffer calsequestrin (CSQN) can cause CTA independently of the steepness of the release-load relationship. In this complementary mechanism, CTA is caused by a beat-to-beat alternation in the number of refractory RyR channels and can occur with or without diastolic SR calcium content alternans depending on pacing conditions and uptake dynamics. The model has unique features, in that it treats a realistic number of spatially distributed and diffusively coupled dyads, each one with a realistic number of RyR channels, and that luminal CSQN buffering and gating is incorporated based on experimental data that characterizes the effect of the conformational state of CSQN on its buffering properties. In addition to reproducing observed features of CTA, this multiscale model is able to describe recent experiments in which CSQN expression levels were genetically altered as well as to reproduce nanoscopic measurements of spark restitution properties. The ability to link microscopic properties of the calcium release units to whole cell behavior makes this model a powerful tool to investigate the arrhythmogenic role of abnormal calcium handling in many pathological settings.

## INTRODUCTION

Calcium transient alternans (CTA) is characterized by beat-to-beat alternations of the peak cytosolic calcium ( $\text{Ca}^{2+}$ ) concentration. The origin of CTA in cardiac myocytes is still not fully understood (1). CTA was originally thought to be caused primarily by alternations of the duration of the cardiac action potential (AP). However, experiments by Chudin et al. (2) demonstrated that the  $\text{Ca}^{2+}$  handling machinery of cardiac myocytes has instabilities of its own that can result in CTA independently of action potential duration (APD) alternans. In these experiments, isolated myocytes were paced with periodic AP waveforms and, at fast pacing rates, CTA was observed to occur. Due to the bidirectional coupling of the membrane voltage and  $\text{Ca}^{2+}$  machinery of the cell (1), instabilities in the  $\text{Ca}^{2+}$  system can lead to APD alternans that has been shown both in mathematical models (3) and experiments (4) to promote fibrillation. Cellular alternans has also been linked to electrocardiographic T-wave alternans (4), which are a strong marker for susceptibility to sudden cardiac arrest (5–7). Elucidating the origin of intracellular CTA has thus become an important problem in excitation-contraction (EC) coupling.

There have been recent attempts (8,9) to construct models of intracellular  $\text{Ca}^{2+}$  dynamics to explain the occurrence of

CTA. These models have incorporated the experimentally observed relationship between the amount of  $\text{Ca}^{2+}$  released from the sarcoplasmic reticulum (SR) and the SR  $\text{Ca}^{2+}$  load. In particular, the steepness of the release-load relationship at high load (10,11) has been proposed as a possible mechanism of CTA (12). (Other factors might coregulate CTA, e.g., the rate of SR  $\text{Ca}^{2+}$  reuptake (1).) Shiferaw et al. (8) incorporated the nonlinear SR release-load relationship in a mathematical model where discrete  $\text{Ca}^{2+}$  release events are spatially averaged over the cell. In the model of Shiferaw et al., the diastolic SR  $\text{Ca}^{2+}$  content displays alternans whenever there is CTA. Although this model partially accounts for observed features of CTA, more recent experimental observations appear inconsistent with a steep SR release-load relationship being the only cause of CTA. In particular, Picht et al. (13) have reported the observation of CTA without diastolic SR load alternans. These observations suggest the possibility of other mechanisms capable of producing CTA without diastolic SR alternans.

The main aim of this article is to identify a new mechanism to explain these observations using a physiologically more realistic model where both luminal gating of SR  $\text{Ca}^{2+}$  release and the spatially distributed nature of unitary  $\text{Ca}^{2+}$  release events are treated explicitly. Although this multiscale myocyte model is used here primarily to study alternans, it could be used more generally to study a wide range of arrhythmogenic mechanisms linked to abnormal  $\text{Ca}^{2+}$  handling.

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## Luminal gating of SR $\text{Ca}^{2+}$ release

A crucial element in modeling EC coupling is the mechanism by which luminal  $\text{Ca}^{2+}$  controls SR release. In the Shiferaw model, it was assumed that the amount of SR  $\text{Ca}^{2+}$  release is a phenomenological nonlinear function of the SR  $\text{Ca}^{2+}$  load which becomes steep for high loads (8). The origin of this nonlinear relationship, however, is not well understood. Calcium release from the SR is regulated by the ryanodine receptors (RyRs), which open upon a rise in the local cytosolic  $\text{Ca}^{2+}$  concentration (14). There is a growing body of experimental evidence showing that luminal  $\text{Ca}^{2+}$  regulates the sensitivity of the RyRs by the interaction of auxiliary proteins (triadin-1/junctin, T/J) with the luminal  $\text{Ca}^{2+}$  buffer calsequestrin (CSQN) (e.g., (15–17)). In particular, Györke et al. (16) have demonstrated the role of CSQN and the T/J complex in regulating the luminal  $\text{Ca}^{2+}$  dependence of the activity of purified RyR channels in lipid bilayers. Additionally, Terentyev et al. (15) have shown that unitary  $\text{Ca}^{2+}$  release events—the so-called  $\text{Ca}^{2+}$  sparks—in transgenic mice with over(under)-expressed CSQN show longer (shorter) refractory periods, thus demonstrating the crucial role of CSQN in modulating the activity of the RyRs. The picture emerging from these experiments is one where, for high luminal  $\text{Ca}^{2+}$  concentration, CSQN is mostly bound to luminal  $\text{Ca}^{2+}$  and unbound from T/J, and the RyR sensitivity to  $\text{Ca}^{2+}$  in the proximal space is higher, i.e., its transition rate from the closed to the open state is higher. On the other hand, for low luminal concentration, CSQN binds to the T/J complex thereby decreasing RyR sensitivity, with a lower transition rate from the closed to the open state.

These experimental observations lead us to test two hypotheses in this article. The first is that the refractoriness of the RyR channels associated with the delay in CSQN unbinding from T/J can cause CTA independently of whether or not the SR load is alternating. The second is that the  $\text{Ca}^{2+}$  concentration dependence of the CSQN binding/unbinding kinetics can explain the steepness of the observed SR release-load relationship at high load. The motivation for testing this second hypothesis is that this nonlinear release-load relationship is well characterized experimentally. It thus provides a valuable benchmark for testing our mathematical model even if, as we conclude here, its steepness at high load may not necessarily be the primary cause of CTA.

## Spatially distributed discrete nature of $\text{Ca}^{2+}$ dynamics

To test these hypotheses theoretically, it is necessary to develop a mathematical model capable of predicting how the gating kinetics of individual  $\text{Ca}^{2+}$  release channels influences the whole cell  $\text{Ca}^{2+}$  transient. This task is made especially arduous by the fact that the whole cell  $\text{Ca}^{2+}$  signal is the summation of discrete release events occurring at a large number of  $\text{Ca}^{2+}$  release units (CRUs) spatially distributed

throughout the cell, as postulated by the local control theory of EC coupling (18); a single CRU typically consists of a few L-type channels in close proximity with a much larger number ( $\sim 50$ – $100$ ) of RyRs. The spatially distributed nature of  $\text{Ca}^{2+}$  release is manifest in a number of phenomena ranging from  $\text{Ca}^{2+}$  waves (19–23), to intracellular spatially discordant alternans (i.e., distinct regions of the cell with CTA of opposite phase) (24–26), to triggered activity (19,27,28) associated with spontaneous release of  $\text{Ca}^{2+}$  at high SR load.

Previous discrete models of intracellular  $\text{Ca}^{2+}$  dynamics can be broadly categorized into two classes. The first class includes models that treat the stochastic dynamics of CRUs to varying degrees of physiological realism (29–34), but assume that the CRUs are not diffusively coupled. While these models have shed light on various aspects of both local and whole-cell  $\text{Ca}^{2+}$  dynamics, such as the distribution of spark life time (30) and the gradedness of  $\text{Ca}^{2+}$  release (31–33), respectively, they do not reproduce CTA even when the  $\text{Ca}^{2+}$  cycling dynamics is coupled to a whole set of membrane ionic currents (33). The second class consists of models that explicitly treat the diffusive coupling between CRUs to study spatio-temporally complex behavior such as intracellular  $\text{Ca}^{2+}$  waves (35–41). These models, however, do not generally incorporate a physiologically detailed description of the stochastic dynamics of CRUs and do not couple the  $\text{Ca}^{2+}$  activity to whole cell membrane currents. Hence, they have so far not been used to study phenomena such as CTA that are governed by the bidirectional coupling between the intracellular  $\text{Ca}^{2+}$  activity and the membrane voltage dynamics. In short, none of the existing models that take into account the discrete and spatially distributed nature of  $\text{Ca}^{2+}$  release events are able to faithfully reproduce the generation of  $\text{Ca}^{2+}$  transient alternans with or without SR content alternans (13).

The model developed in this article distinguishes itself from previous models by its attempt to be both physiologically realistic and truly multiscale. Efficient numerical algorithms are developed that allow us to treat for the first time simultaneously (to our knowledge), in a whole cell model, the dynamics of a realistic number of  $\sim 10,000$  diffusively coupled CRUs interacting with  $\text{Ca}^{2+}$  sensitive membrane currents and other whole cell membrane currents, with each CRU having a realistic number of  $\sim 100$  RyR channels. Furthermore, the model takes into account CSQN-mediated luminal gating of RyR sensitivity based on recent experimental data (16,42,43). As we show here, both luminal gating of RyR sensitivity and diffusive coupling between CRUs turn out to be crucial ingredients for the genesis of CTA which have not been considered in previous modeling studies of CTA, and the model can reproduce CTA both with and without SR content alternans in qualitative agreement with experimental observations (13).

## DESCRIPTION OF THE MODEL

In this section we will describe the most important aspects of our multiscale myocyte model. A more detailed description of

the equations and model parameters is given in the Appendix. In Elementary  $\text{Ca}^{2+}$  Release Unit Structure we describe the intracellular compartments and the various currents in our model. In Cell Architecture we briefly describe the geometry of ventricular myocytes, which we use to realistically implement our spatially extended model. Finally, in the subsections Luminal Gating and Luminal Buffering we present a new mathematical formulation of CSQN-mediated luminal gating and buffering that takes into account the transition from monomeric to dimeric forms of CSQN with increasing luminal free  $\text{Ca}^{2+}$  concentration.

### Elementary $\text{Ca}^{2+}$ release unit structure

Excitation-contraction coupling, the process by which cardiac myocytes transform the membrane depolarization signal into cell contraction, is a complex process that spans multiple scales (44,45). Calcium ions ( $\text{Ca}^{2+}$ ) enter the cell upon membrane depolarization, triggering discrete  $\text{Ca}^{2+}$  release events at the elementary CRUs of the SR, an intracellular store whose primary function is the sequestration and release of intracellular  $\text{Ca}^{2+}$ . These  $\text{Ca}^{2+}$  sparks (46,47) are highly localized in space ( $\sim 1 \mu\text{m}$ ) and time ( $\sim 20 \text{ ms}$ ). The SR is a sac-like structure which forms a spatially dense network of interconnected tubules and cisternae. The SR tubular network is usually referred to as network SR (NSR), while the cisternae are referred to as junctional SR (JSR). These cisternae are localized in close proximity to the T-tubules, cell membrane invaginations that form also a dense structure in ventricular myocytes (48). The cisternae are usually called dyads, and the space between a dyad and the sarcolemma is referred to as the dyadic junction. We will also refer to the dyadic junction as the proximal space of the corresponding dyad.

Upon cell membrane depolarization, L-type  $\text{Ca}^{2+}$  channels colocalized with the dyads (49,50) open, allowing  $\text{Ca}^{2+}$  entry and increasing the  $\text{Ca}^{2+}$  concentration in the proximal space. The pronounced rise in  $\text{Ca}^{2+}$  concentration in the proximal space is sensed by a cluster of RyR channels in the dyad, estimated to have from 50 to 250 RyR channels (14,50), which open and release a larger amount of  $\text{Ca}^{2+}$  from the JSR. The resulting localized rise in  $\text{Ca}^{2+}$  concentration, a  $\text{Ca}^{2+}$  spark, constitutes an elementary release event. Calcium sparks are typically independent of each other, although  $\text{Ca}^{2+}$  released from a spark can trigger neighboring sparks in conditions of elevated SR  $\text{Ca}^{2+}$  load (19). After  $\text{Ca}^{2+}$  is released from a dyad, it diffuses to the submembrane space. The submembrane space is the region of the cell which is close to the sarcolemma, and in which concentrations are sensed by the ion channels corresponding to  $\text{Ca}^{2+}$  sensitive membrane currents, including the  $\text{Ca}^{2+}$  current  $I_{\text{Ca}}$  and the sodium-calcium exchanger current  $I_{\text{NaCa}}$  (8,44,51). Subsequently,  $\text{Ca}^{2+}$  diffuses to the myoplasm, where it activates the cell's contractile machinery. In the myoplasm, the SERCA pump reuptakes a fraction of the  $\text{Ca}^{2+}$  into the NSR, while most of the rest is extruded from the cell by the sodium-calcium exchanger

current in the submembrane space. For a comprehensive description of EC coupling, see Bers (44).

Our goal is to construct a myocyte multiscale model from which whole-cell dynamical behavior, in particular CTA and spatial  $\text{Ca}^{2+}$  dynamics, can be obtained from the properties of the microscopic  $\text{Ca}^{2+}$  handling dynamics described above. Due to the relatively slow (compared to voltage)  $\text{Ca}^{2+}$  diffusion, such microscopic processes are essentially local in nature. Until now, physiologically detailed models of intracellular  $\text{Ca}^{2+}$  dynamics (e.g., (8,32,51)) have assumed global  $\text{Ca}^{2+}$  compartments (for example, Greenstein and Winslow (32) assumes a global cytoplasmic compartment). A novel feature of our model is that the entirety of  $\text{Ca}^{2+}$  dynamics takes place in local units which comprise a dyadic junction and its vicinity, including its proximal and submembrane spaces, the region of myoplasm close to it, its corresponding junctional SR, and the region of the nonjunctional SR in the vicinity of the dyadic junction. Furthermore, as we will discuss in more detail in the next section, these basic units are coupled by  $\text{Ca}^{2+}$  diffusion, thereby endowing our model with spatial organization.

The elements playing a role in the process described above, and which constitute an elementary release unit in our model, are schematically depicted in Fig. 1 *a*. As described in the next subsection, our cell model consists of  $N$  coupled units, and we will denote the quantities associated to unit  $n$  by a superindex ( $n$ ), where  $n = 1, 2, \dots, N$ . In every unit we have a local cytosol, a submembrane space, a proximal space, a JSR space, and an NSR space with volumes  $v_i^{(n)}$ ,  $v_s^{(n)}$ ,  $v_p^{(n)}$ ,  $v_{\text{JSR}}^{(n)}$ , and  $v_{\text{NSR}}^{(n)}$  and  $\text{Ca}^{2+}$  concentrations  $c_i^{(n)}$ ,  $c_s^{(n)}$ ,  $c_p^{(n)}$ ,  $c_{\text{JSR}}^{(n)}$ , and  $c_{\text{NSR}}^{(n)}$ , respectively. The experimentally measured cytoplasmic  $\text{Ca}^{2+}$  concentration  $c_i$  corresponds in our model to the average

$$c_i = \frac{1}{N} \sum_{n=1}^N c_i^{(n)}. \quad (1)$$

In the following, we describe the  $\text{Ca}^{2+}$  currents in a CRU, which are depicted in Fig. 1 *b*. Within each unit, there are diffusive currents between the proximal space and the submembrane space  $I_{\text{ps}}^{(n)}$ , between the submembrane space and the local myoplasm  $I_{\text{si}}^{(n)}$ , and a refilling diffusive current between the local part of the NSR and the SR,  $I_{\text{tr}}^{(n)}$ . Additionally, we have a local uptake current  $I_{\text{up}}^{(n)}$  and a release current  $I_r^{(n)}$  from the JSR into the proximal space. There are diffusive currents between the submembrane spaces, local NSRs, and local myoplasmic spaces of spatially adjacent units. There are several L-type channels per dyadic junction, which release  $\text{Ca}^{2+}$  into the proximal space. For each channel, we have implemented a fully stochastic version of the Markov model of Mahajan et al. (52). The total number of RyR channels per dyadic junction has been estimated to vary between 50 and 250 (14,50), and we will consider 100 RyR channels. The number of L-type  $\text{Ca}^{2+}$  channels has been estimated to be between 1/4 and 1/10 of the number of RyR channels. However, for simplicity, we will consider only four L-type channels per dyad as in Bondarenko et al. (53). A larger

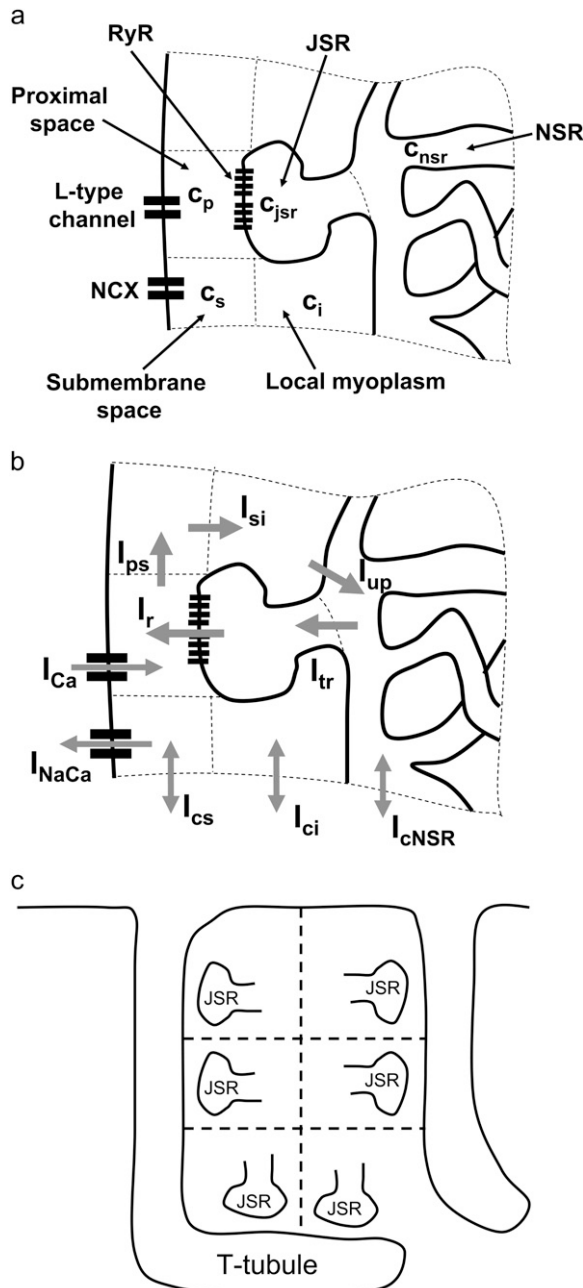


FIGURE 1 (a) Structure of an elementary  $\text{Ca}^{2+}$  release unit (CRU). Each unit consists of local cytosolic, submembrane, proximal, JSR, and NSR compartments. L-type  $\text{Ca}^{2+}$  channels release  $\text{Ca}^{2+}$  into the proximal space and the sodium-calcium exchange current acts on the submembrane space. (b) Currents in a CRU. The arrows illustrate the average direction of  $\text{Ca}^{2+}$  flow. The form of the currents is detailed in the Appendix. (c) Schematic illustration of the heterogeneous orientation of the T-tubules inside the myocyte. The dyadic junctions are oriented along the T-tubules.

number of LCC channels per dyad would result in a more heterogeneous  $I_{\text{Ca}}$  current amplitude per CRU, since different numbers of channels could open within some narrow time interval. Rather than increasing the number of L-type  $\text{Ca}^{2+}$  channels per CRU, we will include the effect of heterogeneity in the  $I_{\text{Ca}}$  signal by considering variations in the proximal

space volume (54) (which affects the concentration of  $\text{Ca}^{2+}$  in the proximal space) as will be discussed in the next section.

The dynamics of RyR gating is a central feature of our model and will be discussed in a separate section. Finally, in each submembrane space we have electrogenic Na-Ca exchangers. The magnitude of the current in each submembrane space, determined by the number of exchangers per unit, was chosen so that the resulting whole cell sodium-calcium exchange current is physiological. The form of the currents and more details can be found in the Appendix.

## Cell architecture

In the past subsection we described  $\text{Ca}^{2+}$  dynamics in a single CRU. A ventricular myocyte has between 10,000 and 100,000 such units (14,32,44). In this subsection we discuss the spatial architecture of the diffusively coupled units that constitute our myocyte model.

The cell membrane has invaginations (the T-tubules) at the so-called Z-planes, which are spaced approximately every  $2 \mu\text{m}$  in the longitudinal direction of the cell (14). Inside the cell, the T-tubules form a complex network which is concentrated on the Z-planes (48). The invagination of the cell membrane into this tubular network allows a more uniform influx of  $\text{Ca}^{2+}$  into the cell resulting in a more even recruitment of  $\text{Ca}^{2+}$  sparks throughout the cell volume. Accordingly, most dyadic junctions are localized in close proximity to the T-tubules. As a result of this arrangement, CRUs are distributed uniformly throughout the cell volume in a regular pattern. For simplicity, we will model the myocyte as a three-dimensional grid, where each element of the grid corresponds to a CRU. We will assume that heterogeneities in the spatial configuration of the elementary units can be spatially averaged and incorporated into effective diffusion coefficients as will be discussed below.

The average separation of the CRUs has been measured using various techniques. Chen-Izu et al. (14) report an average transverse separation of  $1.05 \mu\text{m}$  and an average longitudinal separation of  $1.87 \mu\text{m}$ , while Parker et al. (55) report  $0.76 \mu\text{m}$  in the transverse direction and  $1.8 \mu\text{m}$  in the longitudinal direction; we will use a longitudinal separation of  $l_L = 1.84 \mu\text{m}$  and a transverse separation of  $l_T = 0.9 \mu\text{m}$ .

The total number of CRUs can be estimated to be  $N \sim v_{\text{cell}} / (l_L l_T^2)$ . By processing several confocal microscopy images, Satoh et al. (56) estimate the average volume of ventricular myocytes to be  $\sim v_{\text{cell}} \approx 30 \text{ pL}$ , which yields  $N \sim 20,000$ . The three-dimensional grid by which we represent the myocyte has  $n_x = 65$  units in the longitudinal direction, and  $n_y = 27, n_z = 11$  units in the two transverse directions (see Table 1). These dimensions were chosen to approximately reproduce the observed myocyte shape aspect ratio of  $140 \mu\text{m} \times 30 \mu\text{m} \times 12 \mu\text{m}$  (56). Note that the units themselves are longer in the longitudinal direction.

A large fraction of the cell volume is occupied by myofilaments and mitochondria (44). Therefore, we will consider an

**TABLE 1 Cell architecture parameters**

Parameter	Description	Units
$N$	Number of CRUs	19,305
$n_x$	Number of CRUs along length	65
$n_y$	Number of CRUs along width	27
$n_z$	Number of CRUs along height	11

effective cytosolic volume of  $v_{\text{cyto}} = 10.0 \text{ pl} < v_{\text{cell}}$ . This cytosolic volume is composed from the individual cytosolic volumes of the release units. For unit  $n$ , we denote the corresponding volumes by a superindex ( $n$ ). We take  $v_i^{(n)} = v_{\text{cyto}}/N = 20v_s^{(n)} = 20v_{\text{NSR}}^{(n)}$ , and for the JSR we use  $v_{\text{JSR}}^{(n)} = 2 \times 10^{-2} \mu\text{m}^3$  (57) (see Table 2).

The characteristics of the elements constituting the CRUs have considerable variation, from the number of RyR channels per unit (14,50) to the number of L-type channels colocalized with the unit (53). While, in principle, most of the parameters of a release unit should have random variation, for simplicity we only consider heterogeneity in the volume of the proximal space. The combined effect of the variation in all the parameters will ultimately affect the excitability of the release unit to  $\text{Ca}^{2+}$  triggering and the amount of  $\text{Ca}^{2+}$  released during a spark. The proximal space volume affects the sensitivity of the release unit since it directly affects the maximum  $\text{Ca}^{2+}$  concentration reached in that unit in response to  $\text{Ca}^{2+}$  entry through LCC channels, and thus whether or not  $\text{Ca}^{2+}$ -sensitive RyR channels are activated. We will consider a different volume for each proximal space,  $v_p^{(n)} = 1.26 \times 10^{-3} \mu\text{m}^3 \times (1 + r^{(n)})$ , where  $r^{(n)}$  is a random number drawn from a Gaussian distribution with mean 0 and standard deviation 0.3, truncated so as to restrict  $r^{(n)}$  to the interval  $(-0.8, 0.8)$ . We note that we choose the random numbers  $r^{(n)}$  at the beginning of a simulation and do not change them subsequently. The average proximal space volume,  $1.26 \times 10^{-3} \mu\text{m}^3$ , is obtained by considering the proximal space as a cylinder of radius 200 nm and height 10 nm (30). We find that this proximal space volume heterogeneity modulates the high voltage region of the graded release curve, consistent with the results of Bondarenko et al. (53), and that, in turn, this modulation decreases the propensity for CTA (see Results).

In our model, spatially adjacent units are coupled by diffusive currents. As detailed in the Appendix, the currents have the form  $(c_1 - c_2)/\tau$ , where  $c_1$  and  $c_2$  are the  $\text{Ca}^{2+}$  concentrations in the compartments of the adjacent spaces which are

coupled by the current, and  $\tau$  is a characteristic diffusion time which depends on the effective  $\text{Ca}^{2+}$  diffusion coefficient and the geometrical arrangement of the units. Because the contractile elements of the cell are arranged longitudinally,  $\text{Ca}^{2+}$  has more difficulty diffusing in the transverse direction of the cell, as has been demonstrated by comparing the wave propagation speed in the transverse and longitudinal directions (22). Commonly used values for the transverse and longitudinal diffusion coefficients are  $D^T = 0.15 \mu\text{m}^2/\text{ms}$  and  $D^L = 0.30 \mu\text{m}^2/\text{ms}$  (35). The characteristic time for diffusion between compartments of two neighboring CRUs can be roughly estimated to be on the order of  $\tau \sim l^2/D$ , where  $l$  and  $D$  is the separation between CRUs and diffusion coefficient, respectively. For the cytosolic and NSR compartments we use the corresponding longitudinal and transverse CRU separations  $l_L$  and  $l_T$ , while for the submembrane compartments we use a separation 200-nm shorter to account for the width of the dyadic junction.

Several factors complicate the crude estimation of the times for diffusion  $\tau$  described above. First, the submembrane spaces should be coupled along the T-tubules: when the T-tubules are oriented in the transverse direction, adjacent units in the transverse direction should be coupled, and similarly when the T-tubules are oriented in the longitudinal direction. Although the T-tubules invaginate in the transverse direction, they branch out and bend inside the cell so that both orientations are possible (see Fig. 1 of Soeller and Cannell (48) and the schematic representation in Fig. 1 c). We assume that the majority of the submembrane coupling occurs between transversally adjacent units, but that some longitudinally adjacent units are also coupled. We incorporate this assumption into the model by increasing the longitudinal submembrane diffusive timescales, representing a weaker effective coupling in the longitudinal direction due to the smaller number of units coupled in that direction. In addition, experiments show that CRUs are intercalated between Z-planes in the periphery of the myocyte (14), and modeling studies suggest that these intercalated CRUs increase the probability of wave initiation (35). We include the effect of these extra CRUs by decreasing the longitudinal time constants by a factor of 2. As the previous discussion suggests, the values that one obtains from the estimates above are very crude; thus, rather than adopting a fixed set of values throughout the article, we explore the effect of varying the diffusive timescales in Diffusive Coupling (see later). We find that stronger diffusive coupling promotes the onset of CTA. The values of the diffusive timescales that we use for the other plots are somewhat smaller than what one obtains by the estimates above and are listed later in Table 9. We have checked that the speed of  $\text{Ca}^{2+}$  waves in our model ( $\sim 85 \mu\text{m/s}$ ) agrees with the experimental values ( $50 \sim 150 \mu\text{m/s}$ ) (58,59).

We emphasize that, besides the voltage, there is no global variable that couples the release units. When the action potential is clamped, the only influence of one unit upon another is through the diffusion of  $\text{Ca}^{2+}$ .

**TABLE 2 CRU effective volumes**

Parameter	Description	Units
$v_i$	Local cytosolic volume	$0.5 \mu\text{m}^3$
$v_s$	Local submembrane space volume	$0.025 \mu\text{m}^3$
$v_{\text{JSR}}$	Local JSR volume	$0.02 \mu\text{m}^3$
$v_{\text{NSR}}$	Local NSR volume	$0.025 \mu\text{m}^3$
$\langle v_p \rangle$	Proximal space volume average*	$0.00126 \mu\text{m}^3$

\*The proximal volume  $v_p$  is different for each CRU.

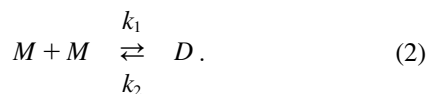
## Luminal gating

In this section, we will construct a model for luminal gating of the RyR channels which regulate  $\text{Ca}^{2+}$  release from the SR. It is well known that SR  $\text{Ca}^{2+}$  content affects, in a nonlinear way, the amount of  $\text{Ca}^{2+}$  that is released from the SR upon stimulation. In particular, the experiments in the literature (10,11) show that, for low SR content ( $<500 \mu\text{M}$ ),  $\text{Ca}^{2+}$  release is quite small, while it increases markedly as SR  $\text{Ca}^{2+}$  content is increased. Some models of intracellular CTA (8) are based on this observation.

There is experimental evidence that an important mechanism by which luminal  $\text{Ca}^{2+}$  controls SR  $\text{Ca}^{2+}$  release is by modulation of the activity of RyR channels through interactions with the  $\text{Ca}^{2+}$  luminal buffer CSQN. CSQN is the main buffering protein in the SR, and is localized mainly in the JSR (60). In experiments in vitro, it was shown (42) that CSQN forms dimers as the  $\text{Ca}^{2+}$  concentration rises above  $\sim 500 \mu\text{M}$ , and forms higher order polymers at concentrations of a few millimolar. Lipid bilayer experiments (16) have shown that the activity of RyR channels is modulated by the interaction of CSQN with the auxiliary proteins triadin-1 and junctin (T/J): the open probability of purified RyR channels sharply increases upon addition of T/J, and it decreases again when CSQN is present. Furthermore, when  $\text{Ca}^{2+}$  concentration on the luminal side is increased from  $20 \mu\text{M}$  to  $5 \text{ mM}$  (binding to and polymerizing CSQN), the open probability of the channels increases again. In addition, in vivo experiments (15) show that transgenic mice with over(under)-expressed CSQN result in  $\text{Ca}^{2+}$  sparks with longer (shorter) refractory periods.

These experiments suggest that CSQN acts as a luminal  $\text{Ca}^{2+}$  sensor and regulates RyR channel activity by attaching to the T/J complex of the RyR channels. Although there is so far no comparable data on the physiological range of  $c_{\text{JSR}}$ , these observations indicate that polymerization of CSQN might modulate RyR sensitivity by inhibiting the attachment of CSQN to the T/J complex. We will construct a simple model of CSQN luminal gating by assuming that only CSQN monomers can bind to and inhibit RyR channels. Although this might be an oversimplification, we believe that such a model captures the general features of CSQN modulation of RyR activity.

In vitro experiments in Park et al. (42) show that a sharp transition from CSQN monomers to dimers occurs as  $c_{\text{JSR}}$  is increased past  $\sim 500 \mu\text{M}$ . Denoting CSQN monomers by  $M$  and CSQN dimers by  $D$ , and assuming rates  $k_1(c_{\text{JSR}})$  and  $k_2(c_{\text{JSR}})$  of dimerization and monomerization dependent on luminal  $\text{Ca}^{2+}$  concentration  $c_{\text{JSR}}$ , respectively, we have the dimerization reaction



The transition rates are believed to be fast (61), so we assume that the dimers and monomers are in equilibrium, obtaining

$[M]^2\rho = [D]$ , where  $\rho \equiv k_1/k_2$  is assumed to depend on the luminal  $\text{Ca}^{2+}$  concentration. The monomer and dimer concentrations are related to the total CSQN concentration,  $B_{\text{CSQN}}$  (assuming no higher polymeric forms, which is approximately true at physiological concentrations (42)), by  $B_{\text{CSQN}} = [M] + 2[D]$ . Note that  $B_{\text{CSQN}}$  refers to the concentration of CSQN molecules, not of CSQN binding sites; the binding capacity of CSQN will be included below. Solving these equations for the relative monomer concentration  $\hat{M} \equiv [M]/B_{\text{CSQN}}$ , we obtain

$$\hat{M} = ((1 + 8\rho B_{\text{CSQN}})^{1/2} - 1)/(4\rho B_{\text{CSQN}}), \quad (3)$$

an expression for  $\hat{M}$  in terms of  $B_{\text{CSQN}}$  and the parameter  $\rho$ . To account for the small range of luminal  $\text{Ca}^{2+}$  concentration over which the monomer-to-dimer transition occurs as observed in Park et al. (42), we will take  $\rho$  to have a sharp dependence on luminal  $\text{Ca}^{2+}$ ,  $c_{\text{JSR}}$ , with parameters chosen so that  $\hat{M}$  decreases from 1 to 0 at  $c_{\text{JSR}} \sim 500 \mu\text{M}$ . For simplicity, we take a sigmoid function

$$\rho(c_{\text{JSR}}) = \frac{\rho_{\infty} c_{\text{JSR}}^h}{K^h + c_{\text{JSR}}^h}, \quad (4)$$

where  $\rho_{\infty} = 5 \times 10^3$ ,  $K = 850 \mu\text{M}$ , and  $h = 23$ . We remark that we choose a large Hill coefficient so as to reproduce the experimentally observed sharp transition from monomers to dimers as a function of  $c_{\text{JSR}}$ . In addition to our hypothesis that only the monomeric form of CSQN can effectively bind to the T/J complex, we assume that the RyR channels have four distinct states, schematically depicted in Fig. 2 *a*: closed CSQN-unbound (1), open CSQN-unbound (2), open CSQN-bound (3), and closed CSQN-bound (4). The transition rates from the closed to the open states are assumed to depend on the proximal  $\text{Ca}^{2+}$  concentration for the CSQN-unbound and CSQN-bound states, respectively, as

$$k_{12}(c_p) = K_u c_p^2, \quad (5)$$

$$k_{43}(c_p) = K_b c_p^2, \quad (6)$$

where  $K_u > K_b$  implies that CSQN-unbound channels have a higher open probability than the CSQN-bound channels. The transition rates from the closed to open states are assumed to be constant and independent of whether the RyR channel is CSQN-bound or CSQN-unbound:

$$k_{21} = k_{34} = \tau_c^{-1}. \quad (7)$$

The transition rates between the closed CSQN-unbound and the closed CSQN-bound states are obtained by assuming simple kinetic rates for CSQN monomer binding and unbinding to the T/J complex, i.e., we assume for simplicity that the binding rate depends linearly on the monomer concentration  $[M]$ , and that the unbinding rate is constant:

$$k_{14} = \tau_b^{-1} [M]/B_{\text{CSQN}}^0, \quad (8)$$

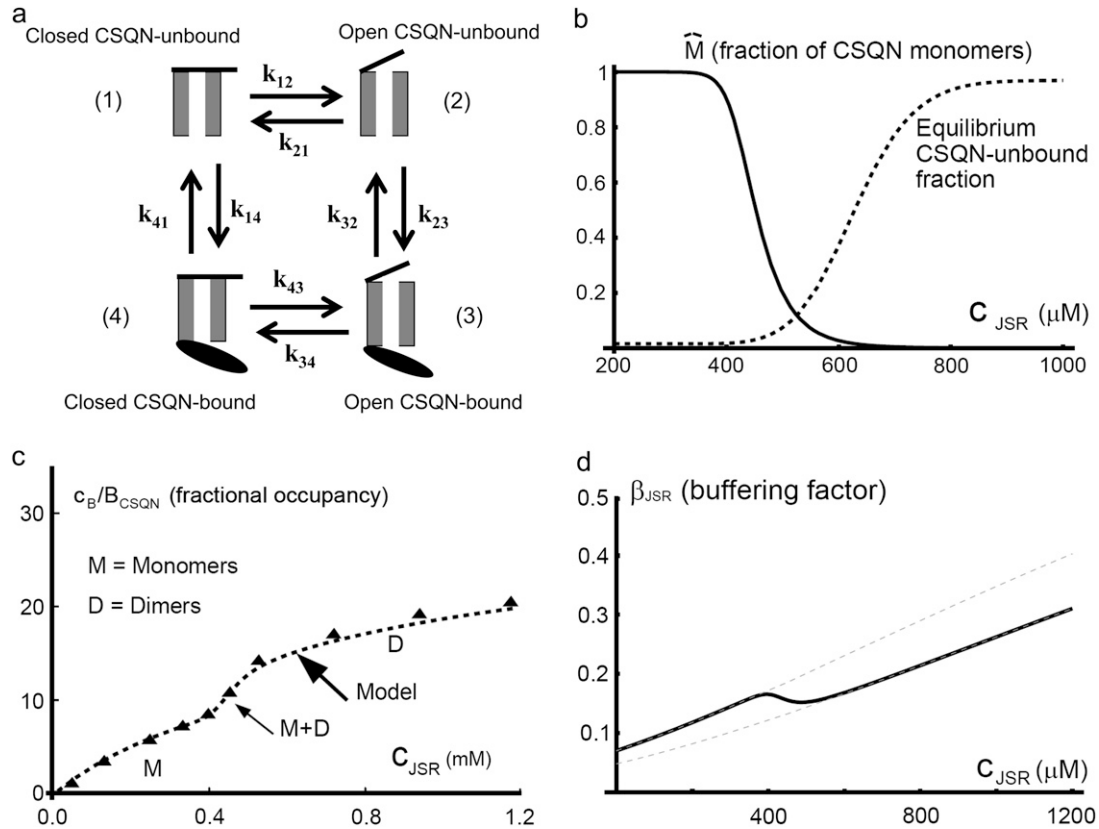


FIGURE 2 (a) Schematic representation of the four-state Markov model representing the four possible states of a RyR channel: closed CSQN-unbound (1), open CSQN-unbound (2), open CSQN-bound (3), and closed CSQN-bound (4). (b) Fraction of monomers  $\hat{M}$  (solid line), and equilibrium value of the CSQN-unbound fraction of RyR channels (dashed line), as a function of the JSR free  $\text{Ca}^{2+}$  concentration  $c_{\text{JSR}}$ . (c) Fractional occupancy,  $c_B/B_{\text{CSQN}}$ , as a function of free  $\text{Ca}^{2+}$ ,  $c_{\text{JSR}}$ . Experimental values for cardiac CSQN from Park et al. (42) (triangles, solid line), and corresponding theoretical prediction from Eq. 14 (dashed line). (d) Buffering factor  $\beta_{\text{JSR}}(c_{\text{JSR}})$  (thick solid line), and the buffering factors that result from assuming that  $n = n_M$  or  $n = n_D$  (thin dashed lines) as a function of  $c_{\text{JSR}}$ .

$$k_{41} = \tau_u^{-1}. \quad (9)$$

where  $\tau_u$  and  $\tau_b$  are the characteristic unbinding (slow) and binding (fast) times, and  $B_{\text{CSQN}}^0 = 400 \mu\text{M}$  is the normal CSQN concentration and guarantees that  $k_{14} \rightarrow \tau_b^{-1}$  for low  $c_{\text{JSR}}$ . We will choose these constants based on experimentally measured spark restitution curves. The transition rate from the open CSQN-unbound to the open CSQN-bound state is taken to be the same as the one from the closed CSQN-unbound to the closed CSQN-bound state,  $k_{23} = k_{14}$ . Finally, to satisfy detailed balance,  $k_{12}k_{23}k_{34}k_{41} = k_{14}k_{43}k_{32}k_{21}$ , we set the transition rate from the open CSQN-bound to the open CSQN-unbound to be

$$k_{32} = k_{41}k_{12}/k_{43}. \quad (10)$$

The value of the parameters can be found later in Table 10 and further details can be found in the Appendix.

We will model each dyad as having 100 RyR channels. Each channel evolves stochastically, independent of the other channels in the dyad. However, to avoid keeping track of the  $\sim 20,000 \times 100$  RyR channels in the myocyte, we only keep

track in each dyad of the number of channels that are in each of the four states. As described in the Appendix, these numbers can be updated in each time-step in a way that is equivalent to individually evolving each channel. Therefore we are able to speed up the simulations by a factor of  $\sim 20$  (from 100 RyRs and four L-type channels to the number of RyR channels per state and four L-type channels).

A crucial feature of the model described above is that the transition rate from the CSQN-unbound to the CSQN-bound states depends dynamically on the luminal  $\text{Ca}^{2+}$  concentration through the dependence of the monomer concentration  $[M]$  on  $c_{\text{JSR}}$ . In Fig. 2 b we plot the fraction of monomers  $\hat{M}$  (solid line) and the equilibrium value of the CSQN-unbound fraction of channels (dashed line) that results from the rates in Eqs. 8 and 9 as a function of  $c_{\text{JSR}}$ . Our luminal gating model result in the following qualitative behavior: just before a dyad fires, its luminal  $\text{Ca}^{2+}$  content is high, and therefore the fraction of CSQN-unbound RyR channels is also high (see Fig. 2 b). When a spark occurs, the luminal  $\text{Ca}^{2+}$  decreases, and hence the fraction of monomers increases. The monomers bind to the T/J complex of the RyR channels and the fraction

of CSQN-bound channels increases. These channels have a lower open probability, and therefore the spark terminates shortly thereafter. Subsequently, the JSR refills, and eventually the luminal  $\text{Ca}^{2+}$  concentration becomes high enough that the concentration of monomers decreases. CSQN unbinds from the T/J complex of the RyR channels, and the dyad returns to its original, excitable state. Although previous models have incorporated CSQN modulated luminal gating of RyR channel activity (51,62), a novel feature of our model is that this modulation is incorporated naturally, resulting from the dynamics of CSQN binding and unbinding from the T/J complex. The various rate parameters are chosen so as to reproduce experimental results and can be found in the Appendix. In particular, the characteristic binding and unbinding times,  $\tau_b$  and  $\tau_u$ , determine the spark restitution properties. It has been observed experimentally that after a spark occurs, sparks elicited shortly thereafter in the same dyad have a smaller amplitude (43,63,64), and that the amplitude of these second sparks recovers gradually so that sparks elicited after  $\sim 400$  ms have the same amplitude as the first spark. The possibility that the smaller amplitude of sparks is due to local refilling of the local JSR stores after depletion is not consistent with the nanoscopic measurement of  $\text{Ca}^{2+}$  blinks (local JSR depletions (43)), which show a much faster recovery timescale than local recovery of spark amplitude. The characteristics of the spark amplitude recovery curve discussed above indicate that the binding time constant  $\tau_b$  should be short, while the unbinding time constant  $\tau_u$  should be of the order of  $100 \sim 200$  ms (see Eqs. 8 and 9). We choose these constants so as to reproduce the spark restitution curve in Brochet et al. (43) (Fig. 4), as discussed in Results and detailed later in Table 10.

### Luminal buffering

Most of the  $\text{Ca}^{2+}$  in the JSR is buffered by CSQN, which can bind tens of  $\text{Ca}^{2+}$  ions per CSQN monomer. Since in our model luminal gating of the RyR channels is inextricably linked to the conformational state of CSQN, i.e., whether it is in monomer or dimer form, an accurate description of CSQN buffering is necessary to quantitatively model experiments in which CSQN properties are modified (15,65). In what follows, we incorporate experimental data on the binding capacities of monomeric and dimeric CSQN in a model of CSQN buffering which is consistent with our scheme for luminal gating.

In vitro experiments (42) have shown that different polymeric forms of CSQN have different buffering capacities. CSQN in dimeric form can absorb up to  $n_D \sim 30 \text{ Ca}^{2+}$  per CSQN molecule (i.e., 60 per dimer), while CSQN in monomeric form can absorb up to  $n_M \sim 15 \text{ Ca}^{2+}$  (see Fig. 6 of Park et al. (42)). For simplicity, we will assume that the monomeric and dimeric forms have the same affinity for  $\text{Ca}^{2+}$ , and that the only difference in their buffering capacity is the number of buffering sites (the subsequent analysis could be generalized

to account for different affinities). With these assumptions, the rate of change in bound luminal concentration  $c_B$  is given by

$$\dot{c}_B = c_{\text{JSR}} k_{\text{on}} m + c_{\text{JSR}} k_{\text{on}} d - c_B k_{\text{off}}, \quad (11)$$

where  $m$  and  $d$  are the concentration of free buffering sites on the monomer and dimer forms, respectively;  $c_{\text{JSR}}$  is the concentration of free  $\text{Ca}^{2+}$ ; and  $k_{\text{on}}$  and  $k_{\text{off}}$  are the rates of binding and unbinding, respectively. The concentration of monomeric buffering sites is  $\hat{M} n_M B_{\text{CSQN}}$ , and the concentration of occupied monomeric buffering sites is  $m_B = \hat{M} c_B$ , so the concentration of free monomeric sites is

$$m = \hat{M} (n_M B_{\text{CSQN}} - c_B). \quad (12)$$

Similarly, the concentration of free dimeric sites is

$$d = (1 - \hat{M}) (n_D B_{\text{CSQN}} - c_B). \quad (13)$$

Inserting these expressions into Eq. 11 and assuming that CSQN buffering is fast (51), we can solve for the steady-state solution to obtain the bound  $\text{Ca}^{2+}$  concentration as a function of the free  $\text{Ca}^{2+}$  concentration  $c_{\text{JSR}}$

$$c_B = \frac{B_{\text{CSQN}} n(c_{\text{JSR}}) c_{\text{JSR}}}{K_C + c_{\text{JSR}}}, \quad (14)$$

where

$$n(c_{\text{JSR}}) = \hat{M} n_M + (1 - \hat{M}) n_D, \quad (15)$$

and  $K_C = k_{\text{off}}/k_{\text{on}}$ . Fig. 2 c (from Park et al. (42)) shows the experimentally measured amount of  $\text{Ca}^{2+}$  bound,  $c_B$ , over the total concentration of CSQN as a function of  $c_{\text{JSR}}$  (triangles) and the theoretical value obtained from Eq. 14 (thick dashed line).

Differentiating the total  $\text{Ca}^{2+}$  concentration in the JSR,

$$c_T = c_{\text{JSR}} + c_B = c_{\text{JSR}} + \frac{B_{\text{CSQN}} n(c_{\text{JSR}}) c_{\text{JSR}}}{K_C + c_{\text{JSR}}}, \quad (16)$$

with respect to time, we obtain

$$\dot{c}_{\text{JSR}} = \beta(c_{\text{JSR}}) \dot{c}_T, \quad (17)$$

where

$$\beta(c) = \left( 1 + \frac{K_C B_{\text{CSQN}} n(c) + \partial_c n(c) (c K_C + c^2)}{(K_C + c)^2} \right)^{-1}. \quad (18)$$

Note that if the buffering capacities of dimers and monomers are assumed equal,  $\partial_c n = 0$ , we recover the rapid buffering approximation expression (66), with a monotonous increase of  $\beta(c)$  with  $c$ . In Fig. 2 d we plot the buffering factor  $\beta(c)$  (thick solid line) and the buffering factors that result from assuming that  $n = n_M$  (upper dashed line) or  $n = n_D$  (lower dashed line) as a function of  $c$ . The buffering factor  $\beta(c)$  has a marked dip at the monomer-dimer transition, suggesting that, as  $c$  increases past the monomer-dimer transition, the number of new CSQN buffering sites on dimers can bind a substantial amount of  $\text{Ca}^{2+}$ .



Before proceeding to Results, we summarize the main novel aspects of our model:

Our model is the first physiologically detailed model of a  $\text{Ca}^{2+}$  cycling in ventricular myocytes simulating a realistic number of diffusively coupled dyads. Such model will not only allow us to determine the effect of microscopic release characteristics on macroscopic variables, but also opens the way to study spatial intracellular  $\text{Ca}^{2+}$  dynamics in the context of a physiologically detailed model.

Our model treats dynamically the modulation of RyR channels by luminal  $\text{Ca}^{2+}$  through CSQN, and integrates the luminal gating with a model of luminal buffering that takes into account experimental evidence of the variation of CSQN buffering capacity due to  $\text{Ca}^{2+}$ -induced conformational changes. The dynamical delay in the gating mechanism provides a novel mechanism for alternans.

## RESULTS

The excitation-contraction machinery of ventricular myocytes is based on the interplay between the  $I_{\text{Ca}}$  current and release of  $\text{Ca}^{2+}$  from the SR. This interaction has been characterized by important experiments, which we have used to benchmark our model. In particular, our model reproduces graded release of  $\text{Ca}^{2+}$ , the graded dependence of the SR release current on  $I_{\text{Ca}}$ , discussed in Graded Release. In Dyadic Junction Refractoriness we analyze the recovery of the release properties of dyadic junctions after release. Another important benchmark is the SR release-load relationship, which is discussed in its own subsection. In the subsections Action Potential Clamp Pacing and Calcium Transient Alternans we pace our model with an action potential clamp and study the resulting  $\text{Ca}^{2+}$  transients, relating the origin of the CTA that we observe to alternations in the fraction of CSQN-unbound RyR channels. In Analysis of Alternans Dynamics we study theoretically the conditions necessary for the genesis of CTA. Finally, in the section Effects of Microscopic Parameters on Alternans, we describe the effect that various microscopic parameters have on the onset of whole cell CTA.

### Graded release

One of the most important characteristics of the excitation-contraction coupling machinery of the cell is the graded release of  $\text{Ca}^{2+}$  from the SR in response to varying stimuli strength. The interaction of the  $I_{\text{Ca}}$  current with  $\text{Ca}^{2+}$  release from the SR was clearly demonstrated in the experiments of Wier et al. (67), who showed that, when depolarizing rat ventricular myocytes to various clamp pulse voltages, both the peak  $I_{\text{Ca}}$  current and the peak SR  $\text{Ca}^{2+}$  release exhibit a bell-shaped curve when plotted as a function of the clamp voltage, with the SR release curve displaced  $\sim 10$  mV toward lower voltages. To reproduce such a curve, we depolarize the

simulated myocyte from  $-80$  mV to various step potentials, and measure the average  $I_{\text{Ca}}$  (solid line) and SR release (dashed line) currents during the first 20 ms after depolarization. The initial concentration of free luminal  $\text{Ca}^{2+}$  is  $650 \mu\text{M}$ . Our model reproduces the experimental observations, as can be observed in Fig. 3. Fig. 3 *a* shows the experimental result from Wier et al. (67) averaged over many cells, while Fig. 3 *b* shows our simulation results.

We have found, in agreement with Bondarenko et al. (53), that heterogeneity in the release unit parameters modulates graded release. However, as opposed to heterogeneity in the number of L-type  $\text{Ca}^{2+}$  channels considered in Bondarenko et al. (53), our model has heterogeneity in the proximal space volume, as was discussed in Elementary  $\text{Ca}^{2+}$  Release Unit Structure (above). For comparison, we show in Fig. 3 *c* the graded release curve corresponding to no heterogeneity. The

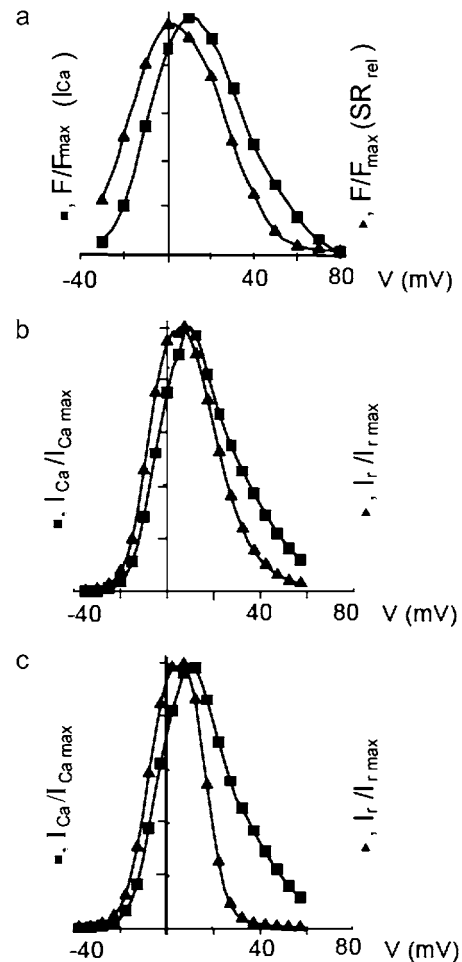


FIGURE 3 (a) Peak L-type  $\text{Ca}^{2+}$  current (squares, rightmost curves) and peak SR release current (triangles, leftmost curves) measured in Wier et al. (67) upon depolarization to various test voltages (horizontal axis). (b) Same quantities obtained with our model, including heterogeneity in the proximal volumes, i.e.,  $v_p^{(n)} = 1.26 \times 10^{-2} \mu\text{m}^3(1 + r^{(n)})$ , where  $r^{(n)}$  is a random number, different for each CRU. (c) Same as in panel b, but with no heterogeneity in the proximal volumes, i.e.,  $v_p^{(n)} = 1.26 \times 10^{-2} \mu\text{m}^3$ .

results show that proximal space volume heterogeneity modulates the high voltage region of the graded release curves, as discussed in Rovetti et al. (54). In this region, the decay of the curve is due to a decreasing driving force, which is compensated by a distribution in the sensitivity of the release units due to heterogeneity. On the negative voltage side, on the other hand, the decay of the curve is caused by a decreasing open probability of the L-type channels, which is not affected by the distribution of the proximal space volumes. We note that in Bondarenko et al. (53) heterogeneity was introduced by multiplying a single channel  $I_{Ca}$  in each CRU by a random number, rather than by considering the addition of a random number of independent L-type  $Ca^{2+}$  channels in each CRU. These approaches are not necessarily equivalent since the L-type channels might fire at different times, thus not significantly increasing the probability of initiating a spark.

### Dyadic junction refractoriness

After  $Ca^{2+}$ -induced  $Ca^{2+}$  release, subsequent release is of smaller magnitude if insufficient time is given for recovery. This recovery from refractoriness, which is a crucial characteristic of the excitation-contraction coupling machinery of the cell, is observed both at the level of whole cell  $Ca^{2+}$  transients (20), and at the level of spark magnitude and firing probability at single dyads (43,63,64). In our luminal gating model, there are two factors that can affect local recovery from refractoriness: 1), refilling of the local JSR; and 2), slow unbinding of CSQN from the T/J complex in the JSR. Sobie et al. (63,64) examined repetitive sparks elicited by addition of 50 nM Ryanodine to a quiescent rat ventricular myocyte. It was found that the recovery of spark amplitude is much faster than the recovery of spark triggering probability; in addition, nanoscopic imaging of local JSR depletions (i.e., blinks) (43) show that local JSR refilling is not the limiting factor governing local recovery from release. In our model, the limiting factor is the recovery from CSQN-binding, which is controlled by a slow rate of CSQN unbinding from the T/J complex,  $\tau_u = 125$  ms. The binding time  $\tau_b$  is set by the observation that the spark triggering probability immediately decreases after a spark is elicited, which implies that the rate of CSQN binding to the T/J complex should be fast; we set  $\tau_b = 2$  ms.

To measure the recovery properties of sparks, we simulate a myocyte held at a potential of 10 mV and collect pairs of sparks occurring in the same CRU during 1 s. During this time,  $c_{JSR}$  averaged over all units had a nearly constant value of  $\sim 800$   $\mu$ M. In the top panel of Fig. 4 we show the amplitude of the second spark normalized by the amplitude of the first spark as a function of the time elapsed between the two sparks for low (3 mM, squares), normal (10 mM, triangles), and high (30 mM, diamonds) concentration of CSQN buffering sites. The figure shows that our model reproduces the observed properties of local recovery from  $Ca^{2+}$  induced  $Ca^{2+}$  release and the experimental results of Terentyev et al. (15), which

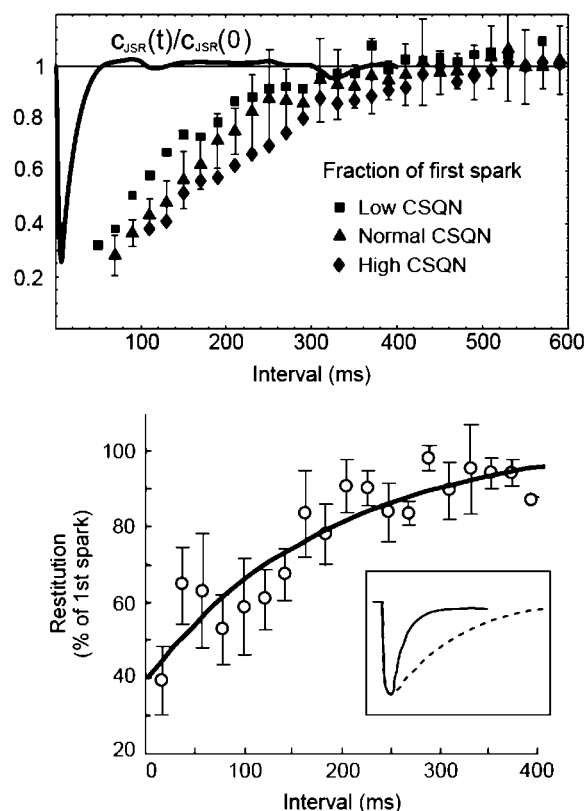


FIGURE 4 (Top) Average amplitude of a spark as a function of the time elapsed after a previous spark occurred in the same CRU for low (squares), normal (triangles, solid line), and high (diamonds) CSQN concentration. The points were grouped in 20-ms intervals; the error bars are the standard deviation in each group for normal CSQN. The horizontal dashed line indicates full spark amplitude recovery. In qualitative agreement with experiments, higher CSQN prolongs the spark refractory period. The solid line is a typical normalized JSR depletion curve,  $c_{JSR}(t)/c_{JSR}(0)$ , showing that the recovery from refractoriness is not associated with local JSR refilling. (Bottom) Experimental results from Brochet et al. (43).

show that the refractory period in individual CRUs increases when CSQN is overexpressed in transgenic mice.

To verify that the spark refractoriness is caused by recovery in the availability of RyR channels and not by refilling of the local JSR stores, we plot in the top panel of Fig. 4 also a typical JSR depletion curve, i.e.,  $c_{JSR}(t)/c_{JSR}(0)$  as a function of time for a single CRU (solid line). As the figure shows, the JSR stores refill much faster than the CRUs recover their original excitability. This result agrees with the experimental observations of Brochet et al. (43), where JSR refilling and spark restitution properties were measured simultaneously, in particular with Fig. 4 B of Brochet et al. (43), reproduced here in the bottom panel of Fig. 4.

### SR release-load relationship

The effect of luminal  $Ca^{2+}$  on the  $Ca^{2+}$  release properties of the SR is often referred to as the SR release-load relationship.

In Shannon and Bers (11), the amount of  $\text{Ca}^{2+}$  released after a voltage clamp step was measured in isolated rabbit ventricular myocytes as a function of the total SR  $\text{Ca}^{2+}$  load (see also (10)). For values of the total SR load  $c_{\text{SRT}} < \sim 90 \mu\text{mol/l}$  cytosol, there was almost no  $\text{Ca}^{2+}$  release, while the release increased sharply between 90 and 120  $\mu\text{mol/l}$  cytosol. In terms of the free  $\text{Ca}^{2+}$  concentration (after accounting for buffering), the nonlinearity was found in these experiments to occur for concentrations larger than  $c_{\text{SR}} \sim 500 \mu\text{M}$ , approximately the same concentration at which our model exhibits the sharp transition from mostly CSQN-bound RyR channels to mostly CSQN-unbound RyR channels as a function of  $c_{\text{SR}}$  observed in Fig. 2 *b*. Since the CSQN-unbound channels have a higher open probability, this sharp increase in the fraction of CSQN-unbound channels results in a steep SR release-load relationship. We remark that the model leading to Fig. 2 *b* was based on the experimentally observed nonlinearity of the CSQN buffering capacity, an independent observation which gives support to the hypothesis that CSQN-mediated gating mechanisms underlie the nonlinearity of the SR release-load relationship. The SR depletion is defined as (11)

$$V(t) = \begin{cases} V_{\min} + (V_{\max} - V_{\min}) \sqrt{1 - \left(\frac{t - mT}{xT}\right)^2} & mT \leq t \leq mT + xT, \\ V_{\min} & mT + xT < t < (m+1)T, \end{cases} \quad (20)$$

$$\text{SR depletion} = \frac{(c_{\text{SRT}}^0 - c_{\text{SRT}}^{\min})}{c_{\text{SRT}}^0} \times 100, \quad (19)$$

where  $c_{\text{SRT}}^0$  is the SR load just before release and  $c_{\text{SRT}}^{\min}$  is the minimum SR load shortly after release. In Fig. 5 we show the SR depletion as a function of  $c_{\text{SR}}$  before release. To vary  $c_{\text{SR}}$  just before release, we start the simulation with a load of 400  $\mu\text{M}$  and the corresponding equilibrium fraction of CSQN-unbound channels (*dashed line* in Fig. 2 *b*) and keep the voltage at  $-80 \text{ mV}$  until the SR refills to the desired value of  $c_{\text{SR}}$ . For loads at  $\sim 400 \mu\text{M}$ , CSQN monomers bind quickly to T/J, so it is reasonable to assume the equilibrium value for the initial fraction of CSQN-unbound channels. We remark,

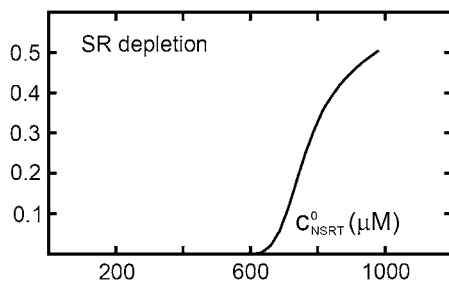


FIGURE 5 SR depletion,  $100 \times ((c_{\text{SRT}}^0 - c_{\text{SRT}}^{\min})/c_{\text{SRT}}^0)$ , as a function of the initial free  $\text{Ca}^{2+}$  SR load,  $c_{\text{SR}}^0$ . A strong nonlinear increase in release for loads larger than  $c_{\text{SR}}^0 \sim 600 \mu\text{M}$  is observed.

however, that the unbinding time is slow and the resulting SR release curve may depend on the number of channels that become CSQN-unbound before the voltage step. When the desired SR load is reached,  $\text{Ca}^{2+}$  release is triggered by a step depolarization from  $-80 \text{ mV}$  to  $10 \text{ mV}$ . The results in Fig. 5 are in qualitative agreement with experiments.

### Action potential clamp pacing

The  $\text{Ca}^{2+}$  cycling machinery of cardiac myocytes and its membrane potential are often thought of as separate dynamical systems coupled by the  $\text{Ca}^{2+}$ -dependent currents (1). The fact that the  $\text{Ca}^{2+}$  system can have instabilities of its own was clearly demonstrated by the experiments of Chudin et al. (2), who paced isolated myocytes with periodic voltage-clamp waveforms and studied the dynamics of the resulting CTA. Since our principal aim is to construct a model of the  $\text{Ca}^{2+}$  cycling machinery of the cell that explains the origin of calcium-induced alternans, we will follow their approach and study the behavior of our model under a periodic voltage-clamp.

Following Shiferaw et al. (8), we will model the action-potential clamp used in Chudin's experiments as

where  $V_{\max} = 15 \text{ mV}$ ,  $V_{\min} = -80 \text{ mV}$ ,  $T$  is the pacing period, and  $x = \text{APD}/T$ , where APD is the action potential duration. As in Shiferaw et al. (8), we take the phenomenological expression  $x(T) = a/(a + T)$  with  $a = 2/3$ , which represents the observed shortening of the APD with decreasing pacing cycle length. In Fig. 6 *a* we plot the averaged cytosolic  $\text{Ca}^{2+}$  concentration  $c_i$  as a function of time that results from pacing our model with a period  $T = 400 \text{ ms}$ . In Fig. 6 *b* we plot the calcium current  $I_{\text{Ca}}$ , and in Fig. 6 *c* we plot the sodium-calcium exchanger current  $I_{\text{NaCa}}$  as a function of time. We remark that the  $I_{\text{Ca}}$  results from the summation of single channel currents determined by independent Markov models at each L-type channel.

### Calcium transient alternans

The fact that single dyadic junctions have a refractory period determined by the slow unbinding of CSQN from the T/J complex allows the possibility of release alternans at the level of a single dyadic junction. Given the stochasticity of the release at the microscopic level, the question arises of whether or not these microscopic alternans can coherently result in macroscopic, whole cell CTA. This is a fundamental problem when one considers the question of CTA in the context of local release; in single pool models, by definition, an instability immediately results in whole cell CTA. In the top panel

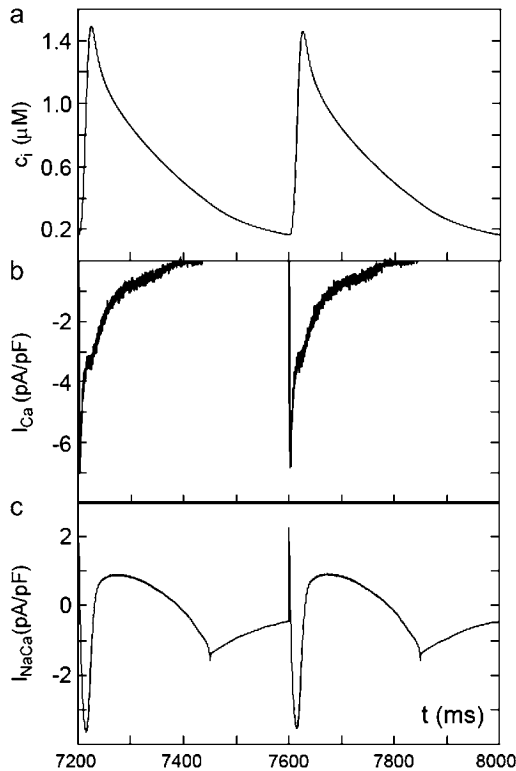


FIGURE 6 (a) Averaged cytosolic  $\text{Ca}^{2+}$  concentration  $c_i$ ; (b) calcium current  $I_{\text{Ca}}$ ; and (c) sodium-calcium exchanger current  $I_{\text{NaCa}}$  as a function of time for a pacing period  $T = 400$  ms showing no CTA.

of Fig. 7 we plot the averaged cytosolic  $\text{Ca}^{2+}$  concentration  $c_i$  (Eq. 1) as a function of time for a pacing period of  $T = 220$  ms. In the bottom panel we show the proximal space  $\text{Ca}^{2+}$  concentration  $c_p$  as a function of time for a transversal line of dyads across the myocyte, indexed in the vertical axis. This panel shows that, even though there is well-defined whole cell CTA, individual dyads do not necessarily reflect this, since, in the time interval shown, there are some dyads firing only in the beats with large  $c_i$ , some firing every beat, and some firing irregularly, examples of which are indicated with the horizontal arrows and marked as *a*, *b*, and *c*, respectively. The vertical arrows in the top panel indicate the time of pacing. In Fig. 8 *a* we show the maximum values of the averaged  $\text{Ca}^{2+}$  cytosolic concentration during steady state pacing for different pacing periods. A bifurcation to CTA appears when the pacing period is decreased at  $< 300$  ms. As we will see in the next subsection, the onset to CTA can be affected by various microscopic parameters. Further insight into the mechanism of alternans can be gained by plotting both the total free SR content, defined as

$$c_{\text{SR}} = \frac{1}{N} \sum_n \frac{v_{\text{JSR}} c_{\text{JSR}}^{(n)} + v_{\text{JSR}} c_{\text{NSR}}^{(n)}}{v_{\text{JSR}} + v_{\text{NSR}}}, \quad (21)$$

and the total number of CSQN-unbound RyR channels as a function of time. In Fig. 8 *b* we show the SR  $\text{Ca}^{2+}$  content  $c_{\text{SR}}$  (dashed line, mM) and the fraction of CSQN-unbound RyR

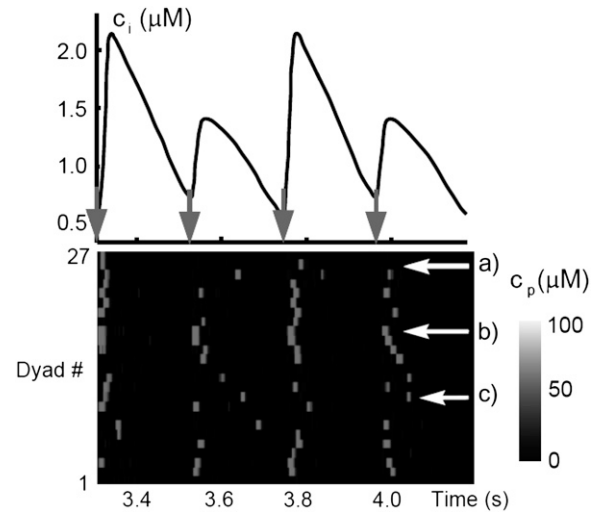


FIGURE 7 (Top) Averaged cytosolic  $\text{Ca}^{2+}$  concentration  $c_i$  as a function of time for a pacing period  $T = 220$  ms showing CTA. (Bottom) Proximal  $\text{Ca}^{2+}$  concentration along a transversal line (27 CRUs) showing how the CTA results from noisy individual signals. In the interval shown, there are dyads firing only in the beats with large  $c_i$ , every beat, and irregularly, examples of which are indicated with the horizontal arrows and marked as *a*, *b*, and *c*, respectively.

channels in the myocyte,  $f$  (solid line), as a function of time. Since the CSQN-unbound RyR channels have a higher open probability, they are important in determining the sensitivity of the release units, and, consequently, the size of the  $\text{Ca}^{2+}$  transient. In the figure we see that, for the parameters used, both the diastolic values of  $f$  and  $c_{\text{SR}}$  alternate. This is not unexpected since, after a large release, uptake by the NSR can be insufficient to bring the SR content back to its original level by the next beat. On the other hand,  $f$  is dependent on the SR content. Therefore, alternations in  $f$  are likely to promote alternations in  $c_{\text{SR}}$ , and vice versa. However, experiments show that it is possible to have CTA without diastolic alternations in  $c_{\text{SR}}$ . Picht et al. (13) measured  $c_{\text{SR}}$  during CTA in rabbit ventricular myocytes and observed that, in some of the cells, CTA was not accompanied by alternations in the diastolic value of  $c_{\text{SR}}$ . In our model, the recovery of RyR channels from CSQN binding provides a mechanism for the genesis of CTA independent of alternations in  $c_{\text{NSR}}$ . In particular, if the timescale of CSQN-unbinding is slow enough compared with the NSR uptake timescale, alternations in  $f$  can occur without alternations in  $c_{\text{NSR}}$ . In Fig. 8 *c* we show the free SR content  $c_{\text{SR}}$  in mM (dashed line) and the fraction of CSQN-unbound RyR channels in the myocyte,  $f$  (solid line), as a function of time for a simulation in which the uptake parameters were modified (the modified parameters are  $v_{\text{up}} = 0.9 \mu\text{M ms}^{-1}$ ,  $K_{\text{NSR}} = 1200 \mu\text{M}$ ,  $K_i = 0.123 \mu\text{M}$ ,  $\xi = 0.1$ , and  $\tau_u = 270$  ms). We remark that in the experiments of Picht et al. (13), this type of alternans was observed only in a fraction of the myocytes. While  $f$  shows clear diastolic alternans,  $c_{\text{SR}}$  does not show significant diastolic alternations. As the period is decreased, small alternans develop in the

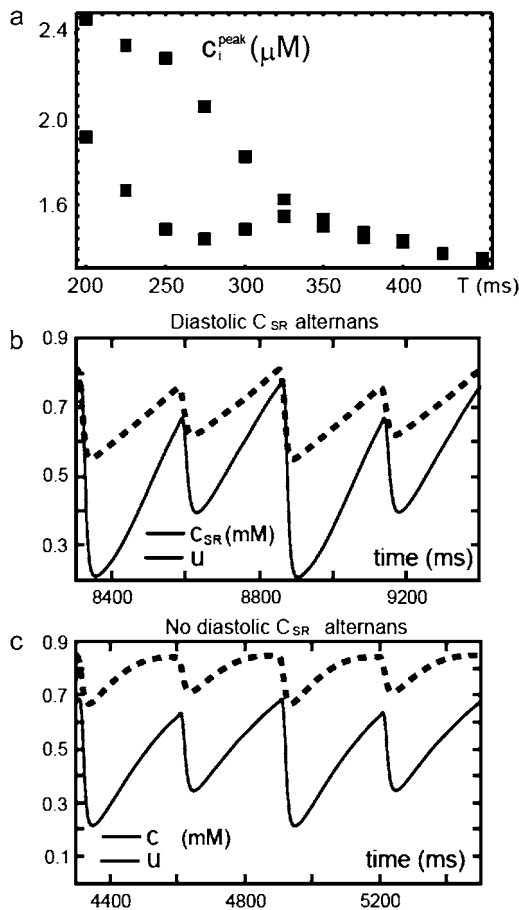


FIGURE 8 (a) Peak averaged cytosolic  $\text{Ca}^{2+}$  concentration  $c_i$  on alternate beats during steady-state pacing as a function of pacing period  $T$ . A transition to CTA occurs as the pacing period is decreased. (b) Free SR content in mM (dashed line) and the fraction of CSQN-unbound RyR channels in the myocyte,  $f$  (solid line) as a function of time for the original parameters. SR release alternans are accompanied by diastolic SR alternans. (c) Free SR content  $c_{\text{SR}}$  in mM (dashed line) and the fraction of CSQN-unbound RyR channels in the myocyte with modified parameters,  $f$  (solid line) as a function of time. Although there are SR release alternans, no significant diastolic SR alternans are present (see discussion in text).

diastolic SR content, as shown in Fig. 9, but they remain small when compared with the alternations in the fraction of unbound channels. This shows that alternations in the number of available (i.e., CSQN-unbound) RyR channels can generate CTA without significant diastolic alternations in SR content.

### Analysis of alternans dynamics

In this subsection we investigate theoretically the relative contributions of the steep SR release-load relationship and the recovery of RyR channels from CSQN binding to the genesis of CTA. In particular, we discuss the conditions required for CTA without diastolic SR content alternans. Our analysis is a direct extension of an iterative map analysis previously developed in the literature (1,8), which studies the onset insta-

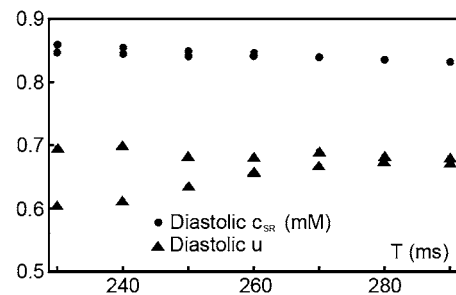


FIGURE 9 Diastolic SR content (circles) and fraction of CSQN-unbound channels  $f$  (triangles) at steady state pacing as a function of the pacing period  $T$  for the modified parameters used in Fig 8 (c).

bilities in the beat-to-beat dynamics of SR load. Here, we will generalize this method by including the beat to beat dynamics of the CSQN-unbound RyR channels. For simplicity, we consider deterministic dynamics in a single CRU. Let  $l_n$  be the SR load and  $f_n$  be the fraction of CSQN-unbound channels just before the  $n^{\text{th}}$  beat. Let  $R(l_n, f_n)$  be the amount by which the SR content decreases just after the  $n^{\text{th}}$  beat as a function of these variables. Let  $U(l_n - R)$  be the amount of  $\text{Ca}^{2+}$  that is uptaken after SR release decreases the load to  $l_n - R$  (see Fig. 10). Furthermore, let  $g(l_n - R)$  be the fraction of RyR channels that remain in the CSQN-unbound state just after the load is decreased to  $l_n - R$  by SR release (since binding is fast, we assume it does not depend on its value before release). We then have the following map relating these quantities at beat  $n + 1$  to their values at beat  $n$ :

$$\begin{aligned} l_{n+1} &= l_n - R(l_n, f_n) + U(l_n - R(l_n, f_n)), \\ f_{n+1} &= g_{\infty} + [g(l_n - R(l_n, f_n)) - g_{\infty}]e^{-T/\tau_u}, \end{aligned} \quad (22)$$

where we have assumed that the recovery of the unbound channels fraction relaxes exponentially to the equilibrium CSQN-unbound fraction  $g_{\infty}$  of RyR channels for large SR load (see Fig. 2 b) with a time constant  $\tau_u$ .

In the absence of alternans, the SR load and fraction of CSQN-unbound channels have the same values every beat,  $l_n = l, f_n = f$ . The conditions for the onset of alternans can be studied by introducing small perturbations from these values,  $l_n = l + \delta l_n, f_n = f + \delta f_n$ , and studying their beat-to-beat evolution. Growth of these perturbations corresponds to the

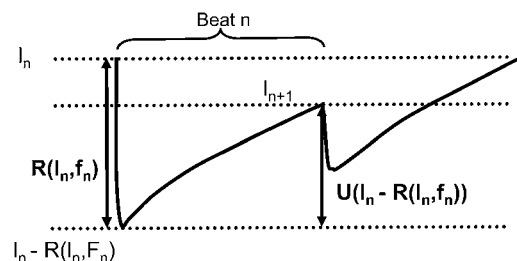


FIGURE 10 Functions used in the map analysis:  $R(l, f)$  is the amount by which the SR content decreases as a function of the load  $l$  and fraction of CSQN-unbound fraction  $f$ , and  $U(l)$  is the amount of uptaken  $\text{Ca}^{2+}$  after the load decreases to  $l$  after release.

development of alternans. Inserting these expressions in Eq. 22 and linearizing, we obtain

$$\begin{pmatrix} \delta l_{n+1} \\ \delta f_{n+1} \end{pmatrix} = \begin{pmatrix} (1 - \partial_l R)(1 + \partial_l U) & -\partial_l R(1 + \partial_l U) \\ \partial_l g(1 - \partial_l R)e^{-T/\tau_u} & -\partial_l g \partial_l R e^{-T/\tau_u} \end{pmatrix} \begin{pmatrix} \delta l_n \\ \delta f_n \end{pmatrix}, \quad (23)$$

where the derivatives are evaluated at the values  $l, f$ . The condition for alternating growth of the perturbations is that the eigenvalue of the matrix in Eq. 23 with the largest magnitude is  $< -1$  (for alternans, this eigenvalue is negative). This condition results in that alternans develop when

$$(1 + \partial_l U)(\partial_l R - 1) + \partial_l R \partial_l g e^{-T/\tau_u} > 1. \quad (24)$$

This equation is the key to understanding the relative contributions of steep SR release-load relationship, uptake, and recovery of CSQN-bound channels. If we can neglect CSQN-mediated effects,  $\partial_l R = 0$ ,  $\partial_l g = 0$ , or  $\tau_u \ll T$ , then we recover the criterion presented in the literature (1,8), that alternans occur only when

$$\partial_l R > \frac{2 + \partial_l U}{1 + \partial_l U}. \quad (25)$$

In this case, alternans develop as a result of a combination of a steep SR release-load relationship (large  $\partial_l R$ ) and poor uptake (small  $\partial_l U$ ; we note that here  $\partial_l U$  is negative since a smaller  $\text{Ca}^{2+}$  load results in more uptake). In the more general case described by Eq. 24, in addition to steep SR release-load relationship and poor uptake, alternans can result from a combination of steep dependence of release on the fraction of CSQN-unbound channels (large  $\partial_l R$ ), large nonlinearity in CSQN binding due to the steep CSQN polymerization as a function of  $c_{\text{JSR}}$  (large  $\partial_l g$ ), or insufficient time to recover from CSQN binding (large  $\tau_u/T$ ). In particular, it might be possible that the first term vanishes,  $(1 + \partial_l U)(\partial_l R - 1) = 0$ , but that the second term is enough to cause alternans by itself,  $\partial_l R \partial_l g e^{-T/\tau_u} > 1$ . This shows that, in principle, alternans can occur independently of a steep SR release-load relationship.

We can use Eq. 25 to justify our claim that in the simulations shown in Fig. 8 *c* the alternations in the fraction of CSQN-unbound channels drive the alternans, rather than the very small, but still present, alternations in the SR content. The drop in the SR content in beats with large release is  $\sim 180 \mu\text{M}$ , and in beats with small release it is  $\sim 130 \mu\text{M}$ , so the difference in SR depletion is  $\sim 50 \mu\text{M}$ . The difference in SR content before release is  $\sim 10 \mu\text{M}$ . Thus, since in this argument we are neglecting the effects of  $f$  on release, we can estimate that  $\partial_l R \approx 5$ . We can also estimate  $\partial_l U$  by noting that a release  $R \approx 180 \mu\text{M}$  is reuptaken so that by the next beat the load is approximately within  $10 \mu\text{M}$  of its previous value  $l_0$ . Therefore, the amount of reuptaken  $\text{Ca}^{2+}$  after the load is reduced to  $l$  is in this case at least  $U \approx 170 \mu\text{M}/180 \mu\text{M} \times (l_0 - l) = 0.94(l_0 - l)$ , and thus we estimate  $\partial_l U \approx -0.94$ . The condition in Eq. 25 is not fulfilled and this shows that the steepness of the SR release-load relationship and uptake alone cannot cause the observed release alternans.

Another important question is what are the conditions to observe release alternans without diastolic SR alternans. We can obtain the relative amplitudes of alternations in diastolic SR content and in the fraction of CSQN-unbound RyR channels by finding the eigenvector of the matrix in Eq. 23 corresponding to alternans, which is

$$\begin{pmatrix} \delta l \\ \delta f \end{pmatrix} = \begin{pmatrix} 1 + \partial_l U \\ \partial_l g e^{-T/\tau_u} \end{pmatrix}. \quad (26)$$

This gives the relative amplitudes of the diastolic SR content and CSQN-unbound fraction of RyR channel alternans. We see that, to have no significant SR content alternans with release alternans, we must have  $1 + \partial_l U \ll \partial_l g e^{-T/\tau_u}$ . The left-hand side is small if  $\partial_l U \approx -1$ , which corresponds from Eq. 22 to constant  $l_n$  (no SR alternans). In our simulations, the SR content attains a steady-state value in diastole ( $\partial_l U \approx -0.94$ ), while in the experiments of Picht et al. (13) it has a more triangular shape and does not attain steady state while still reaching the same SR diastolic value. Although in both cases this results in  $\partial_l U \sim -1$ , this qualitative difference indicates that the mechanism by which this is accomplished in both cases is different. Details of the uptake dynamics allowing the uptake for a large release to exactly balance the uptake for a small release without reaching steady state are a separate subject of investigation and are not considered here. Our analysis shows that, independently of how the SR load is kept constant at diastole, CSQN-mediated recovery of the RyR channels from the unbound state can produce release alternans.

As an additional proof of the principle that CTA can be driven by alternations in the fraction of unbound CSQN channels independently of alternations in diastolic SR content, we have modified the parameters so that the NSR content does not fluctuate appreciably (e.g., as if it were highly buffered) and have observed cases in which the diastolic alternations of the SR content are negligible, while the alternations in SR release and in the CSQN-unbound fraction are still large (not shown).

The steep release load relationship (Fig. 5) has been proposed by various studies (8,12) as a mechanism for the genesis of CTA. Our results support the hypothesis that there is another complementary variable, the number of available RyR channels, which, together with the SR content, determines the amount of  $\text{Ca}^{2+}$  released from the SR.

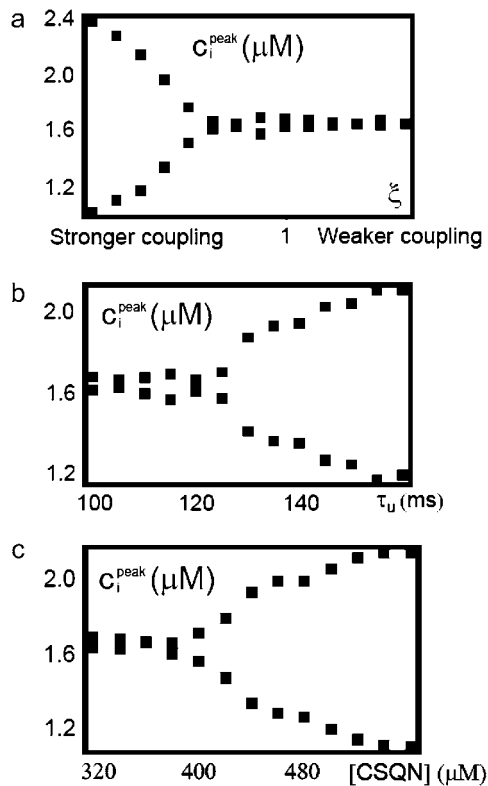
## EFFECT OF MICROSCOPIC PARAMETERS ON ALTERNANS

In this section we will discuss the effect of some microscopic properties of the CRUs on the onset of CTA that our spatially distributed model is uniquely capable of investigating.

### Diffusive coupling

Whole cell CTA results from the summation of the stochastic activity of a large number of CRUs. Even though we have

argued that one can expect beat-to-beat alternations in the release at individual dyads based on their gating dynamics, CRUs are highly stochastic and, as shown in the bottom panel of Fig. 7, they do not necessarily fire in phase with the whole cell CTA. If the CRUs fired independently from each other, one would expect them to fire incoherently, resulting in no CTA. The reason that the release at the level of single dyadic junctions do not dephase stochastically is that neighboring dyadic junctions are coupled diffusively: if, averaged over the myocyte, CRUs are firing in an off-on-off-on pattern resulting in whole cell CTA,  $\text{Ca}^{2+}$  diffusion will increase the probability that an individual CRU will fire in a beat in which the majority of CRUs (and, in particular, a majority of its neighboring CRUs) are firing, thus reinforcing and maintaining the CTA. Without diffusive coupling, the release at individual CRUs would desynchronize, resulting in no whole cell CTA. To test this hypothesis, we varied the diffusive coupling timescales close to the onset of alternans. In Fig. 11 *a* we show the peak values of  $c_i$  during steady-state alternans as



**FIGURE 11** (a) Maximum averaged cytosolic  $\text{Ca}^{2+}$  concentration  $c_i$  on alternate beats during steady-state pacing as a function of the coupling strength scaling factor  $\xi$ . The parameter  $\xi$  scales all the nearest-neighbor diffusive timescales (i.e.,  $\tau \rightarrow \xi\tau$ ). Stronger coupling promotes CTA. (b) Maximum averaged cytosolic  $\text{Ca}^{2+}$  concentration  $c_i$  on alternate beats during steady-state pacing as a function of the unbinding time  $\tau_u$ . Larger unbinding time promotes CTA. (c) Maximum averaged cytosolic  $\text{Ca}^{2+}$  concentration  $c_i$  on alternate beats during steady-state pacing as a function of the concentration of CSQN sites. A larger concentration of CSQN sites promotes CTA. When not varied, the parameter values are  $\tau_u = 125$  ms,  $[\text{CSQN}] = 400$   $\mu\text{M}$ , and  $\xi = 0.4$ .

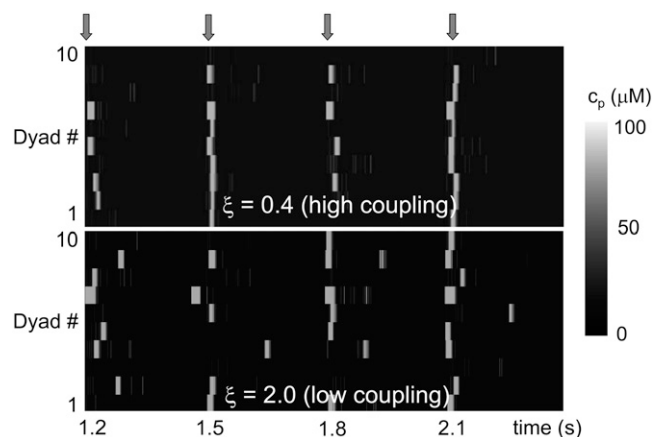
a function of a parameter  $\xi$ , which controls the coupling strength. The parameter  $\xi$  scales all the nearest-neighbor diffusive timescales (i.e.,  $\tau \rightarrow \xi\tau$  for the  $\tau$ -values specified later in Table 9), so that higher  $\xi$  corresponds to weaker coupling. The figure shows that, as argued above, diffusive coupling promotes the onset of alternans. In Fig. 12 we show the firing pattern along a transversal line including 10 CRUs for two values of the diffusive timescales' scaling factor  $\xi$ . Large coupling (Fig. 12, *top*,  $\xi = 0.4$ ) promotes CRU firing synchronization, while for low coupling (Fig. 12, *bottom*,  $\xi = 2.0$ ) CRUs fire more independently.

### Calsequestrin unbinding time

By directly affecting the duration of the CRU recovery from release, the CSQN unbinding timescale  $\tau_u$  has an important effect on the onset of CTA. The timescale of recovery has been estimated to be  $\sim 180$  ms (43). In Fig. 11 *b* we show the peak values of  $c_i$  during steady-state pacing as a function of  $\tau_u$ . The plot shows that CTA is promoted by larger  $\tau_u$ .

### Calsequestrin concentration

In vivo experiments with transgenic mice overexpressing CSQN resulted in CTA at rapid pacing rates for all the transgenic mice in the experiment and no CTA for the control mice (65). The mechanism by which CSQN overexpression might promote alternans was investigated in Terentyev et al. (15), which showed that overexpression of CSQN results in longer CRU refractory periods. This phenomenon is qualitatively reproduced by our model in the top panel of Fig. 4, where we show the CRU restitution curves at various levels of CSQN. One of the novel features of our model is the realistic modeling of the dual role of CSQN as a buffer and as a reg-



**FIGURE 12** Proximal  $\text{Ca}^{2+}$  concentration along a transversal line of 10 CRUs for large (*top*,  $\xi = 0.4$ ) and low diffusive coupling (*bottom*,  $\xi = 2.0$ ). Large coupling promotes CRU firing synchronization, while for low coupling CRUs fire more independently. The arrows on top indicate pacing, with period  $T = 300$  ms.

ulator of RyR activity, which allows us to investigate the role of CSQN in promoting CTA. In Fig. 11 *c* we show the maximum averaged cytosolic  $\text{Ca}^{2+}$  concentration  $c_i$  on alternate beats during steady-state alternans as a function of the concentration of CSQN buffering sites (i.e., sites that bind  $\text{Ca}^{2+}$ ). As observed in experiments, a larger concentration of CSQN promotes CTA.

### Heterogeneity of CRU properties

We have also observed that heterogeneity in the release properties (in our case, the proximal volumes  $v_p^{(n)}$ ) inhibits CTA (data not shown). As we have discussed, heterogeneity smoothes the high-voltage part of the release curve (compare Fig. 3, *b* and *c*). As the membrane voltage  $V$  decreases after the upstroke, it will recruit sparks with a probability roughly determined by the release curve. If this curve decays sharply on the high voltage side, there will be a sharp increase in spark recruitment when the voltage decreases at  $\sim V = 0$  mV. This results in many sparks occurring in a relatively short time interval, thereby facilitating the recruitment of other sparks through diffusion. As discussed above, sustaining whole cell CTA against the tendency of CRUs to fire independently of each other relies on sparks influencing each other through diffusion. On the other hand, if the release curve is smooth, spark recruitment is more gradual, and diffusive coupling is less effective, as there are less sparks being recruited in a given time interval. Thus, it seems that some degree of stochasticity in CRU release properties is beneficial (in the sense that it suppresses whole cell CTA).

### MODEL LIMITATIONS

While our model goes beyond previous efforts in constructing a myocyte  $\text{Ca}^{2+}$  cycling model that captures the spatial dynamics of  $\text{Ca}^{2+}$  release and luminal CSQN-mediated luminal gating, it has some limitations that we now discuss.

Various experimental measurements that we have used to calibrate different parts of our model come from different species: data on CSQN polymerization (Fig. 2 *c*) corresponds to canine CSQN (42), graded release curves (Fig. 3) correspond to rat, and spark restitution curves (Fig. 3) and the L-type  $\text{Ca}^{2+}$  current correspond to rabbit (43,52). Thus, our model at this point can be used to draw conclusions about alternans only in a general and qualitative way. However, it could be adapted to a particular species if enough experimental measurements are available.

As presented here, the model incorporates only elements necessary to describe  $\text{Ca}^{2+}$  cycling and, for simplicity, we have paced the model with a phenomenological action potential waveform (Eq. 20) and used a fixed intracellular sodium concentration at each rate (Eq. 47). We note, however, that the model can be easily coupled to existing electrophysiological models. In preliminary explorations we have

successfully coupled it to the Fox model of ionic membrane currents (68) as modified in Sato et al. (69).

In the experiments of Picht et al. (13), the SR content reaches the same diastolic value without attaining steady state, even though the systolic value alternates. In our model, the SR content reaches the same diastolic value by attaining steady state. While these mechanisms are different, they are related to uptake dynamics and are not central to our argument. Understanding the uptake dynamics that allows systolic SR alternans without diastolic SR alternans is open for future studies.

Our model neglects mechanoelectric feedback effects, which are increasingly being recognized to affect cellular electrophysiology. In particular, stretch-activated channels (70) are not considered, and while the rate of  $\text{Ca}^{2+}$ -Troponin-C binding is tension-dependent (71), we have taken it to be constant. The effect of mechanoelectric feedback on alternans will be explored elsewhere.

The estimates for the effective diffusive timescales later in Table 9 are crude and may oversimplify the complex geometry of the dyadic junctions, including spatial anisotropy of  $\text{Ca}^{2+}$  diffusion and the heterogeneous orientation of neighboring units. Although the effect of simultaneously scaling all these timescales was explored in Fig. 11 *a*, they were not independently varied.

RyR channels can be inactivated by cytosolic  $\text{Ca}^{2+}$  (see (72) and references therein). Our model does not include cytosolic inactivation of the RyR channels. Although, as discussed below, our results suggest CSQN-mediated luminal gating of RyR is the primary mechanism for the genesis of CTA, the possible role of cytosolic inactivation sites was not investigated here.

### DISCUSSION

We have developed a multiscale myocyte model that allows the study of the origin of CTA at the subcellular level. In our model, whole cell CTA results from the coherent summation of irregular alternations of release at the level of single dyadic junctions. The alternans are caused by an alternation in the number of RyR channels that are available to open (CSQN-unbound), which in turn alternate because of the slow recovery associated to the CSQN-unbinding timescale. Importantly, our model admits CTA without significant alternations in diastolic SR  $\text{Ca}^{2+}$  content, as observed in the experiments of Picht et al. (13). However, in most cases, alternations in diastolic SR content accompanied CTA. The reason is that the SR content and the number of available RyR channels are strongly coupled to each other: the rate of CSQN-binding depends strongly on  $c_{\text{JSR}}$  through its effect on the number of CSQN monomers, Eq. 8, while the diastolic SR content depends on the number of available RyR channels in the previous beat (many available RyR channels  $\rightarrow$  large release  $\rightarrow$  low diastolic SR content). Thus, alternations in the diastolic SR content will typically promote alternations in the



number of available RyR channels, and vice versa. However, when the uptake is enough to reduce or eliminate the dependence of the diastolic SR content on the number of available RyR channels in the previous beat, one can observe CTA without significant alternations in diastolic SR  $\text{Ca}^{2+}$  content as in Fig. 8 *c*.

Another important feature of our model is that, for the first time, it simulates a realistic number ( $\sim 20,000$ ) of diffusively coupled, physiologically detailed elementary release units, where each unit has a realistic ( $\sim 100$ ) number of RyR channels. Our numerical method for efficiently simulating the behavior of the RyR channels at a single unit, which tracks only the number of RyR channels in a given state, is equivalent to the independent evolution of the Markov model for each unit. This method, detailed in the Appendix, works because the release current only depends on the number of channels in the open state. The model was numerically simulated using C++ in a desktop Mac with a 2 GHz Intel Core Duo processor. Running a 1-s simulation takes 55 min of computation time.

In heart failure (HF), a characteristic of abnormal  $\text{Ca}^{2+}$  handling is increased RyR leakiness, which results in diminished SR release. In our model, low levels of CSQN result in increased RyR leak since CSQN monomers inhibit the activity of the RyR channels. However, we have found that lower CSQN levels alone inhibit CTA, which are generally associated with HF. Our findings suggest that CTA in HF are caused by other changes, such as increased intracellular sodium concentration (which increases the activity of the exchanger) or downregulation of the SERCA pump. A thorough exploration of the broader changes associated with HF using our model is beyond the scope of this article and will be pursued elsewhere. Besides HF and alternans, our model might also provide a useful framework to study the cellular mechanisms of triggered arrhythmias such as catecholaminergic polymorphic ventricular tachycardia, which has been linked to mutations resulting in reduced CSQN levels (73). In particular, our model shows that such a reduction shortens RyR refractoriness (Fig. 4) and enhances RyR activity, which could promote spontaneous calcium release.

RyR channels can be inactivated by cytosolic  $\text{Ca}^{2+}$ , and it has been suggested that luminal  $\text{Ca}^{2+}$  might regulate the activity of the RyR channels by permeation to cytosolic inactivation sites (see (72) and references therein). Therefore, whether or not cytosolic  $\text{Ca}^{2+}$  inactivation of the RyR channels can also produce CTA, independently of luminal  $\text{Ca}^{2+}$  regulation, is an important question. Our results suggest that inactivation of RyR channels alone is insufficient to produce CTA. This conclusion is supported by the fact that the time-scale for recovery from inactivation is a strongly nonlinear function of  $\text{Ca}^{2+}$  concentration in our model due to the cooperative nature of the CSQN monomer-dimer transition, and that this nonlinearity is essential for CTA to occur. Since there is no obvious analog of this strong nonlinearity in the case where RyR channel activity is only gated by cytosolic  $\text{Ca}^{2+}$ ,

it is unclear whether CTA can occur in this case. In addition, there is compelling experimental evidence for CSQN-mediated luminal  $\text{Ca}^{2+}$  regulation of the RyR channels, and experiments with CSQN-altered transgenic mice show an altered propensity to alternans (65), in agreement with our model results (Fig. 11 *c*). Thus, while we do not rule out a possible contribution of cytosolic  $\text{Ca}^{2+}$  inactivation, we believe that the principal mechanism for the genesis of CTA lies in the luminal  $\text{Ca}^{2+}$  regulation of SR release. The incorporation of cytosolic RyR inactivation to our model is a subject for further research.

Diffusive coupling between adjacent release units was found to promote whole cell CTA by synchronizing the patterns of CRU  $\text{Ca}^{2+}$  release alternans. In addition to being a crucial element in any model studying the nanoscopic origin of CTA, diffusive coupling and spatial structure allow for the future study of other intracellular  $\text{Ca}^{2+}$  dynamics phenomena observed in experiments, like intracellular  $\text{Ca}^{2+}$  waves (19–23), intracellular spatially discordant alternans (24–26), and spontaneously triggered activity (19,27,28). The understanding of such arrhythmogenic phenomena requires a model of intracellular  $\text{Ca}^{2+}$  dynamics that realistically couples the various membrane currents with the  $\text{Ca}^{2+}$  handling machinery of the cell and that takes into account the local, spatially extended nature of the excitation-contraction process. As the resolution of imaging techniques continues to improve, models will be needed which can, with a multiscale approach, address the relation of the measured nanoscopic structures to whole cell signals. Our model allows us to explore the effect of various microscopic variables on whole cell signals, in particular on the origin of CTA. In our model, larger CSQN unbinding time, larger CSQN concentration, and stronger diffusive coupling between neighboring CRUs promote alternans, while heterogeneity in the CRU properties inhibits alternans.

## APPENDIX

### Model variables

Here we describe the basic characteristics of our model. It has  $14N$  variables, where  $N$  is the number of CRUs,  $N = 19,305$ . For a given CRU, these 14 variables are the concentrations in the five CRU compartments; two concentrations of  $\text{Ca}^{2+}$  bound to Troponin-C in the local cytosolic and submembrane compartments; three variables that describe the number of RyR channels in the different states of the four-state RyR Markov model; and four variables that describe the state of each of the four L-type channels in the L-type channel Markov model (see Table 3).

These variables evolve, as will be described below in more detail, by means of  $7N$  diffusively coupled differential equations for the concentrations  $c_s^{(n)}$ ,  $c_p^{(n)}$ ,  $c_{\text{JSR}}^{(n)}$ ,  $c_{\text{NSR}}^{(n)}$ ,  $[\text{CaT}]_i^{(n)}$ ,  $[\text{CaT}]_s^{(n)}$ ,  $4N$  L-type channel Markov models describing the state  $y_j^{(n)}$  of four L-type channels per CRU,  $j = 1, 2, 3, 4$ , and  $N$  Markov models that give the number of RyR channels in the states of the RyR four-state model for each CRU,  $x_i^{(n)}$ ,  $i = 1, 2, 3$ , where  $n = 1, \dots, N$ .

### Dynamics of $\text{Ca}^{2+}$ cycling in an elementary unit

In what follows we describe the  $\text{Ca}^{2+}$  dynamics in the  $n^{\text{th}}$  CRU. To simplify the notation, we will omit here the superscript ( $n$ ) when there is not the

**TABLE 3 Model variables**

Variable	Description	Units
$c_i^{(n)}$	$n^{\text{th}}$ Unit cytosolic $\text{Ca}^{2+}$ concentration	$\mu\text{M}$
$c_s^{(n)}$	$n^{\text{th}}$ Unit submembrane $\text{Ca}^{2+}$ concentration	$\mu\text{M}$
$c_p^{(n)}$	$n^{\text{th}}$ Unit proximal space $\text{Ca}^{2+}$ concentration	$\mu\text{M}$
$c_{\text{NSR}}^{(n)}$	$n^{\text{th}}$ Unit NSR $\text{Ca}^{2+}$ concentration	$\mu\text{M}$
$c_{\text{JSR}}^{(n)}$	$n^{\text{th}}$ Unit JSR $\text{Ca}^{2+}$ concentration	$\mu\text{M}$
$[\text{CaT}]_i^{(n)}$	$n^{\text{th}}$ Unit $\text{Ca}^{2+}$ -Troponin-C cytosolic concentration	$\mu\text{M}$
$[\text{CaT}]_s^{(n)}$	$n^{\text{th}}$ Unit $\text{Ca}^{2+}$ -Troponin-C submembrane concentration	$\mu\text{M}$
$x_i^{(n)}$	Number of RyR channels in $n^{\text{th}}$ unit in state $i$	
$y_j^{(n)}$	State of $j^{\text{th}}$ L-type channel in $n^{\text{th}}$ unit	

possibility of confusion, with the understanding that the currents and concentrations (and the heterogeneous proximal space volume  $v_p$ ) refer to those at the  $n^{\text{th}}$  unit. The dynamics of  $\text{Ca}^{2+}$  cycling in the  $n^{\text{th}}$  CRU is described by the equations

$$\dot{c}_i = \beta_i(c_i) \left( I_{\text{dsi}} \frac{v_s}{v_i} - I_{\text{up}} + I_{\text{leak}} - I_{\text{TCi}} + I_{\text{ci}} \right), \quad (27)$$

$$\dot{c}_s = \beta_s(c_s) \left( I_{\text{dps}} \frac{v_p}{v_s} + I_{\text{NCX}} - I_{\text{dsi}} - I_{\text{TCs}} + I_{\text{cs}} \right), \quad (28)$$

$$\dot{c}_p = \beta_p(c_p) (I_r + I_{\text{Ca}} - I_{\text{dps}}), \quad (29)$$

$$\dot{c}_{\text{NSR}} = \left( (I_{\text{up}} - I_{\text{leak}}) \frac{v_i}{v_{\text{NSR}}} - I_{\text{tr}} \frac{v_{\text{JSR}}}{v_{\text{NSR}}} + I_{\text{cNSR}} \right), \quad (30)$$

$$\dot{c}_{\text{JSR}} = \beta_{\text{JSR}}(c_{\text{JSR}}) \left( I_{\text{tr}} - I_r \frac{v_p}{v_{\text{JSR}}} \right). \quad (31)$$

In what follows we comment on the various currents and terms in Eqs. 27–31. We note that, in general, the results of calibration of mathematical expressions in whole-cell models are not necessarily valid when the same expressions are used in a CRU. The concentrations used in whole cell current expressions are averaged concentrations, while the concentration in CRUs might differ significantly from the average. Since the average of a function is not generally equal to the function of the average, the result of averaging a current over many CRUs is not necessarily the same as using averaged concentrations in a single pool model current. Therefore, in some cases we adjust the whole-cell current parameters that have been calibrated in previous whole cell models while keeping the same functional form.

### Instantaneous buffering functions $\beta(c)$

The term  $\beta_X(c_X)$  accounts for instantaneous buffering in compartment  $X$  using the rapid buffering approximation (66). For example, instantaneous buffering in the cytosolic compartments is given by  $\beta_i(c_i)$  as

$$\beta_i(c_i) = \left[ 1 + \sum_b \frac{B_b K_b}{(c_i + K_b)^2} \right]^{-1}, \quad (32)$$

where the sum is over the instantaneous cytosolic buffers Calmodulin, SR sites, Myosin (Ca), and Myosin (Mg), with buffer dissociation constants  $K_{\text{CAM}}$ ,  $K_{\text{SR}}$ ,  $K_{\text{MCA}}$ , and  $K_{\text{MMg}}$  and total concentration of buffering sites  $B_{\text{CAM}}$ ,  $B_{\text{SR}}$ ,  $B_{\text{MCA}}$ , and  $B_{\text{MMg}}$ , respectively. In addition to Calmodulin, we consider instantaneous buffering in the submembrane space by the subsarcolemmal sites of high affinity (51) with total concentration of sites and dissociation constant  $B_{\text{SLH}}$  and  $K_{\text{SLH}}$ , respectively. As we will discuss below, we assume

that the concentration of the proximal space rapidly equilibrates,  $\dot{c}_p \approx 0$  (see, for example, (30)), so we do not require knowledge of the instantaneous buffers in the proximal space. The values for the buffer parameters are in Table 4, and were obtained from the literature (44,51).

Luminal buffering was described in detail in Luminal Buffering, where it was argued that the expression for  $\beta_{\text{JSR}}(c_{\text{JSR}})$  is

$$\beta(c) = \left( 1 + \frac{K_C B_{\text{CSQN}} n(c) + \partial_c n(c) (c K_C + c^2)}{(K_C + c)^2} \right)^{-1}, \quad (33)$$

where

$$n(c_{\text{JSR}}) = \hat{M}_M + (1 - \hat{M}) n_D, \quad (34)$$

$$\hat{M} = \frac{(1 + 8\rho B_{\text{CSQN}})^{1/2} - 1}{4\rho B_{\text{CSQN}}}, \quad (35)$$

and

$$\rho(c_{\text{JSR}}) = \frac{\rho_{\infty} c_{\text{JSR}}^h}{K^h + c_{\text{JSR}}^h}. \quad (36)$$

The parameters for luminal buffering are in Table 5. We note that  $B_{\text{CSQN}}$  represents concentration of CSQN molecules. Taking an average buffering capacity of 25 sites per CSQN molecule, we would have a concentration of CSQN buffering sites of 10,000  $\mu\text{M}$ . Measurements of CSQN buffering sites concentration range from 3 mM (74) to 14 mM (75). The effect of varying the concentration of CSQN (as has been done with transgenic mice (15,65)) is explored in Fig. 11 c. Shannon and Bers (75) reports a dissociation constant of 630  $\mu\text{M}$  and Mitchell et al. (76) reports a dissociation constant in the range 400–600  $\mu\text{M}$ ; we take  $K_C = 600 \mu\text{M}$ .

### Troponin C dynamic buffering currents, $I_{\text{TCi}}$ and $I_{\text{TCs}}$

These currents describe the rate of change in the concentration of  $\text{Ca}^{2+}$  bound to Troponin C in the cytosolic and submembrane compartments,  $[\text{CaT}]_i$  and  $[\text{CaT}]_s$ . These quantities satisfy

$$\frac{d[\text{CaT}]_i}{dt} = I_{\text{TCi}}, \quad (37)$$

with

$$I_{\text{TCi}} = k_{\text{on}}^{\text{T}} c_i (B_{\text{T}} - [\text{CaT}]_i) - k_{\text{off}}^{\text{T}} [\text{CaT}]_i, \quad (38)$$

and analogous expressions apply for the submembrane compartments, replacing the subscript  $i$  by  $s$ . Here,  $k_{\text{on}}$  and  $k_{\text{off}}$  are the on- and off-rate

**TABLE 4 Buffering parameters**

Parameter	Value	Units
$K_{\text{CAM}}$	7.0	$\mu\text{M}$
$B_{\text{CAM}}$	24.0	$\mu\text{M}$
$K_{\text{SR}}$	0.6	$\mu\text{M}$
$B_{\text{SR}}$	47.0	$\mu\text{M}$
$K_{\text{MCA}}$	0.033	$\mu\text{M}$
$B_{\text{MCA}}$	140.0	$\mu\text{M}$
$K_{\text{MMg}}$	3.64	$\mu\text{M}$
$B_{\text{MMg}}$	140.0	$\mu\text{M}$
$K_{\text{SLH}}$	0.3	$\mu\text{M}$
$B_{\text{SLH}}$	13.4	$\mu\text{M}$
$B_{\text{T}}$	70.0	$\mu\text{M}$
$k_{\text{on}}^{\text{T}}$	0.0327	$(\mu\text{M ms})^{-1}$
$k_{\text{off}}^{\text{T}}$	0.0196	$\text{ms}^{-1}$

**TABLE 5 Luminal buffering parameters**

Parameter	Description	Value
$B_{\text{CSQN}}$	Concentration of CSQN molecules	400 $\mu\text{M}$
$K_{\text{C}}$	Dissociation constant of CSQN	600 $\mu\text{M}$
$n_{\text{M}}$	Buffering capacity of CSQN monomers	15
$n_{\text{D}}$	Buffering capacity of CSQN dimers	35
$\rho_{\infty}$	Asymptotic ratio of dimers to monomers	5000
$K$	Dimerization constant	850 $\mu\text{M}$
$h$	Dimerization exponent (steep)	23

constants for  $\text{Ca}^{2+}$ -Troponin C binding, and  $B_{\text{T}}$  is the total concentration of Troponin C buffering sites.

### Uptake current $I_{\text{up}}$

We use the uptake current from Shannon et al. (51)

$$I_{\text{up}} = v_{\text{up}} \frac{(c_i/K_i)^H - (c_{\text{NSR}}/K_{\text{NSR}})^H}{1 + (c_i/K_i)^H + (c_{\text{NSR}}/K_{\text{NSR}})^H}. \quad (39)$$

See Table 6 for parameters used.

### Leak current $I_{\text{leak}}$

The leak current has the form Shiferaw et al. (8)

$$I_{\text{leak}} = g_{\text{leak}} \frac{(c_{\text{JSR}})^2}{(c_{\text{JSR}})^2 + K_{\text{JSR}}^2} (c_{\text{NSR}} - c_i). \quad (40)$$

See Table 6 for parameters used.

### Sodium-calcium exchange current $I_{\text{NaCa}}$

We used the exchange current as in Shannon et al. (51)

$$I_{\text{NaCa}} = \frac{K_{\text{a}} v_{\text{NaCa}} (e^{\eta z} [\text{Na}]_i^3 [\text{Ca}]_o - e^{(\eta-1)z} [\text{Na}]_o^3 c_s)}{(t1 + t2 + t3)(1 + k_{\text{sat}} e^{(\eta-1)z})}, \quad (41)$$

where

$$t1 = K_{\text{mCai}} [\text{Na}]_o^3 \left[ 1 + \left( \frac{[\text{Na}]_i}{K_{\text{mNai}}} \right)^3 \right], \quad (42)$$

$$t2 = K_{\text{mNao}}^3 c_s \left[ 1 + \left( \frac{c_s}{K_{\text{mCai}}} \right) \right], \quad (43)$$

$$t3 = K_{\text{mCao}} [\text{Na}]_i^3 + [\text{Na}]_i^3 [\text{Ca}]_o + [\text{Na}]_o^3 c_s, \quad (44)$$

**TABLE 6 Uptake and leak current parameters**

Parameter	Value	Units
$v_{\text{up}}$	0.3	$\mu\text{M ms}^{-1}$
$K_i$	0.123	$\mu\text{M}$
$K_{\text{NSR}}$	1700	$\mu\text{M}$
$H$	1.787	
$g_{\text{leak}}$	$1.035 \times 10^{-5}$	$\text{ms}^{-1}$
$K_{\text{JSR}}$	500	$\mu\text{M}$

$$K_{\text{a}} = \left[ 1 + \left( \frac{K_{\text{da}}}{c_s} \right)^3 \right]^{-1}, \quad (45)$$

$$z = \frac{VF}{RT}, \quad (46)$$

and the values of the parameters can be found in Table 7.

The internal sodium concentration  $[\text{Na}^+]_i$  adjusts slowly upon a change in pacing rate, reflecting the new equilibrium between sodium influx due to the sodium current, and sodium extrusion by the sodium-calcium exchanger and the sodium-potassium pump. This process occurs over many beats and is not essential for modeling  $\text{Ca}^{2+}$  dynamics. Therefore, following Shiferaw et al. (8), we will adopt a phenomenological expression for  $[\text{Na}^+]_i$  as a function of the pacing period  $T$ ,

$$[\text{Na}^+]_i = \frac{a}{1 + bT^{1/2}}, \quad (47)$$

where  $a = 78 \text{ mM}$ ,  $b = 10 \text{ s}^{-1/2}$ , and  $T$  is in seconds. These values yield  $[\text{Na}^+]_i \approx 14.2 \text{ mM}$  for  $T = 0.2 \text{ s}$  and  $[\text{Na}^+]_i \approx 7.1 \text{ mM}$  for  $T = 1 \text{ s}$ .

### Calcium current, $I_{\text{Ca}}$

Each CRU is assumed to have four L-type channels,  $l = 1, 2, 3, 4$ . The  $\text{Ca}^{2+}$  flux into the proximal space of the  $n^{\text{th}}$  CRU is given by

$$I_{\text{Ca}} = i_{\text{Ca}} N_{\text{L}}, \quad (48)$$

where the single channel current is (77)

$$i_{\text{Ca}} = 4P_{\text{Ca}} z F \frac{\gamma_i \hat{c}_p e^{2z} - \gamma_o [\text{Ca}^{2+}]_o}{e^{2z} - 1}, \quad (49)$$

$z = VF/(RT)$ , and  $\hat{c}_p = 10^{-3} c_p$ , i.e., the concentration of the proximal space in mM units. The gating of each L-type channel is simulated by stochastically evolving the seven-state Markov model developed in Mahajan et al. (52), with their constants  $\hat{c}_p$ ,  $\hat{c}_p$ , and  $r_2$  modified to 1.5, 0.5, and 6.0, respectively. The value  $N_{\text{L}}$  is the number of channels in the open state,  $1 \leq N_{\text{L}} \leq 4$ . The parameter values are in Table 8.

### Release current, $I_r$

The JSR in each CRU releases  $\text{Ca}^{2+}$  into the proximal space through multiple RyR channels. In our model, each CRU has 100 RyR channels. The release current for the  $n^{\text{th}}$  CRU is given by (30)

$$I_r = J_{\text{max}} P_o (c_{\text{JSR}} - c_p) / v_p, \quad (50)$$

where  $P_o$  is the fraction of RyR channels that are in the open states (states 2 and 3 in Fig. 2a), and  $J_{\text{max}} = 1.47 \times 10^{-2} \mu\text{M}^3 \text{ ms}^{-1}$ . The gating of the RyR channels used to obtain  $P_o$  will be described in the next subsection.

**TABLE 7 Sodium-calcium exchanger current parameters**

Parameter	Value	Units
$v_{\text{NaCa}}$	1.05	$\mu\text{M ms}^{-1}$
$K_{\text{mCai}}$	3.59	$\mu\text{M}$
$K_{\text{mCao}}$	1.3	mM
$K_{\text{mNai}}$	12.3	mM
$K_{\text{mNao}}$	87.5	mM
$K_{\text{da}}$	0.11	$\mu\text{M}$
$K_{\text{sat}}$	0.27	
$\eta$	0.35	
$[\text{Ca}^{2+}]_o$	1.8	mM
$[\text{Na}^+]_o$	136	mM

**TABLE 8 Calcium current parameters**

Parameter	Value	Units
$P_{Ca}$	L-type channel permeability	$11.9 \mu\text{mol C}^{-1} \text{ms}^{-1}$
$\gamma_i, \gamma_o$	Activity coefficient of $\text{Ca}^{2+}$	0.341
$F$	Faraday's constant	$96.5 \text{ C mmol}^{-1}$
$R$	Universal gas constant	$8.314 \text{ J mol}^{-1} \text{K}^{-1}$
$T$	Temperature	308 K
$[\text{Ca}^{2+}]_o$	External calcium concentration	1.8 mM

### Diffusion from proximal space to submembrane space, $I_{dps}$

We use a diffusive current of the form  $I_{dps} = (c_p - c_s)/\tau_p$  with  $\tau_p = 0.022$  ms. This time corresponds to diffusion over a distance of  $\sim 80$  nm, the average distance to travel slightly out of the proximal space, which has a cylindrical shape with radius 200 nm. Since  $\tau_p$  is significantly smaller than other timescales in the problem, we assume, following Hinch (30), that  $c_p$  rapidly equilibrates so that  $\dot{c}_p = 0$ . We remark that, in this approximation, the proximal space instantaneous buffering factor  $\beta_{JSR}$  is not needed. Using Eq. 27 and the expressions for  $I_r$  and  $I_{dps}$  to solve  $\dot{c}_p = 0$  for  $c_p$ , we obtain

$$c_p = \frac{c_s + \tau_p(k_r c_{JSR} - I_{Ca})}{1 + \tau_p J_{\max} P_o / v_p}. \quad (51)$$

### Diffusion from NSR to JSR, $I_{tr}$

The JSR refilling current is given by  $I_{tr} = (c_{NSR} - c_{JSR})/\tau_{tr}$ . From nanoscopic imaging of JSR depletions (43), the translocation time from NSR to JSR has been estimated to be  $\sim 30$  ms. We choose  $\tau_{tr} = 5$  ms, so that when taking into account the buffering factor  $\beta_{JSR} \sim 0.15$  (see Fig. 2 d), the effective refilling time is  $5 \text{ ms}/\beta_{JSR} \sim 30$  ms.

### Nearest-neighbor diffusive currents, $I_{ci}$ , $I_{cs}$ , and $I_{cNSR}$

The diffusive currents couple neighboring dyadic junctions. Here, to avoid confusion, it is necessary to distinguish quantities corresponding to the  $n^{\text{th}}$  CRU with the superindex ( $n$ ). The cytosolic diffusive current is given by

$$I_{ci}^{(n)} = \sum_m \left( \frac{c_i^{(m)} - c_i^{(n)}}{\tau_{mn}} \right), \quad (52)$$

where the sum is over the (at most six) nearest neighbors,  $\tau_{mn} = \tau_i^T$  if units  $m$  and  $n$  lie along one of the transverse directions, and  $\tau_{mn} = \tau_i^L$  if they lie along the longitudinal direction (the values for these timescales are in Table 9). More precisely, the CRUs can be labeled with indices  $1 \leq n_x \leq 65$ ,  $1 \leq n_y \leq 27$ , and  $1 \leq n_z \leq 11$  denoting their position in the three-dimensional grid so that  $(n) \rightarrow (n_x, n_y, n_z)$ , with  $x$  representing the longitudinal direction. Then, for  $1 < n_x < 65$ ,  $1 < n_y < 27$ , and  $1 < n_z < 11$ , we have

$$\begin{aligned} I_{ci}^{(n)} = & (c_i^{(n_x+1, n_y, n_z)} + c_i^{(n_x-1, n_y, n_z)} - 2c_i^{(n_x, n_y, n_z)})(\tau_i^L)^{-1} \\ & + (c_i^{(n_x, n_y+1, n_z)} + c_i^{(n_x, n_y-1, n_z)} - 2c_i^{(n_x, n_y, n_z)})(\tau_i^T)^{-1} \\ & + (c_i^{(n_x, n_y, n_z+1)} + c_i^{(n_x, n_y, n_z-1)} - 2c_i^{(n_x, n_y, n_z)})(\tau_i^T)^{-1}. \end{aligned} \quad (53)$$

When the unit is in the border (i.e.,  $n_x = 1, 65$ ,  $n_y = 1, 27$ , or  $n_z = 1, 11$ ), we modify the previous expression appropriately to impose no-flux boundary conditions.

Similar expressions are used for the submembrane and NSR diffusive currents  $I_{cs}$  and  $I_{cNSR}$ .

**TABLE 9 Diffusive timescales**

Parameter	Description	Units
$\tau_i^T$	Transverse cytosolic*	2.93 ms
$\tau_i^L$	Longitudinal cytosolic*	2.32 ms
$\tau_{NSR}^T$	Transverse NSR*	7.2 ms
$\tau_{NSR}^L$	Longitudinal NSR*	24.0 ms
$\tau_s^T$	Transverse submembrane*	1.42 ms
$\tau_s^L$	Longitudinal submembrane*	3.4 ms
$\tau_{tr}$	JSR refilling time	5.0 ms
$\tau_{ps}$	Proximal to submembrane	0.022 ms
$\tau_{si}$	Submembrane to cytoplasm	0.1 ms

\*The effect of varying these quantities is shown in Fig. 11.

### Diffusion from submembrane to myoplasm, $I_{dsi}$

We use a diffusive current of the form  $I_{dsi} = (c_s - c_i)/\tau_{si}$  with  $\tau_{si} = 0.1$  ms, estimated as the time it takes for a  $\text{Ca}^{2+}$  ion to diffuse  $\sim 200$  nm into the cytosolic compartment. Since this timescale is fast, we assume instantaneous equilibration,  $\dot{c}_s = 0$ , obtaining from Eq. 28

$$\begin{aligned} c_s^{(n)} = & \left( c_p^{(n)} v_p^n v_s^{-1} \tau_{ps}^{-1} + I_{NaCa}^{(n)} + c_i^{(n)} \tau_{si}^{-1} - I_{TCs}^{(n)} \right. \\ & \left. + \sum_m c_s^{(m)} \tau_{mn}^{-1} \right) / (\tau_{ps}^{-1} + \tau_{si}^{-1} + \tau_{nm}^{-1}). \end{aligned} \quad (54)$$

### RyR gating Markov model

As described in Luminal Gating, we consider a four-state Markov model with the following states: closed CSQN-unbound (1), open CSQN-unbound (2), open CSQN-bound (3), and closed CSQN-bound (4). The rates of transition from state  $i$  to state  $j$ ,  $k_{ij}$ , are

$$\begin{aligned} k_{12} &= K_u c_p^2, \\ k_{14} &= \hat{M}(c_p) \tau_b^{-1} B_{CSQN} / B_{CSQN}^0, \\ k_{21} &= \tau_c^{-1}, \\ k_{23} &= \hat{M}(c_p) \tau_b^{-1} B_{CSQN} / B_{CSQN}^0, \\ k_{43} &= K_b c_p^2, \\ k_{41} &= \tau_u^{-1}, \\ k_{34} &= \tau_c^{-1}, \\ k_{32} &= k_{41} k_{12} / k_{43}, \end{aligned} \quad (55)$$

where  $\hat{M}$  is given by Eq. 35 and the parameters are in Table 10. We note that  $B_{CSQN} / B_{CSQN}^0$  is only different from 1 when we modify CSQN concentration.

A direct simulation of the stochastic evolution of each one of the  $100 \times 20,000 = 2 \times 10^6$  RyR channels in the myocyte would be computationally

**TABLE 10 RyR Markov model parameters**

Parameter	Description	Value
$K_u$	CSQN-unbound opening rate	$3.8 \times 10^{-4} \mu\text{M}^{-2} \text{ms}^{-1}$
$K_b$	CSQN-bound opening rate	$5 \times 10^{-5} \mu\text{M}^{-2} \text{ms}^{-1}$
$\tau_u$	CSQN unbinding timescale	125.00 ms
$\tau_b$	CSQN binding timescale	5.0 ms
$\tau_c$	RyR closing timescale	1.0 ms
$B_{CSQN}^0$	Normal CSQN concentration	400 $\mu\text{M}$

prohibitive, since diffusive coupling between adjacent CRUs requires the simultaneous processing of RyR gating dynamics. To reduce the computation time to reasonable levels, we do not simulate each individual channel in a given CRU, but rather keep track of the number of channels in a given CRU that are in each state. The number of states in the  $n^{\text{th}}$  dyad in the open CSQN-bound (3), open CSQN-unbound (2), and closed CSQN-unbound (1) are denoted by  $x_3^{(n)}$ ,  $x_2^{(n)}$ , and  $x_1^{(n)}$ , respectively (the number of RyRs in the closed CSQN-bound state is  $x_4^{(n)} = 100 - x_1^{(n)} - x_2^{(n)} - x_3^{(n)}$ ). Henceforth we will omit the superscript  $(n)$ . The release current  $I_r$  depends only on the fraction of states in the open states  $P_o = (x_2 + x_3)/100$ , rather than on which particular channels are in each state. Therefore, at each time step we only need to compute the number of channels that make transitions from one state to another. Since we have the probabilities for the transition of an individual channel, the distribution of the number of channels making a transition from state  $j$  can be obtained from a multinomial distribution with the number of trials being the number of RyRs in state  $j$  and the probabilities of success being the probabilities of transition to another state given by the expressions in Eq. 55. We remark that, so far, this is an equivalent mathematical formulation of the process that requires, for the large number of RyR channels we consider, less computational effort. Further approximations allow us to increase the efficiency of the simulation. In practice, the probabilities of transition per unit time are small and we can treat transitions to different states as independent. For example, if at time  $t$  there are  $x_1$  channels in the closed unbound state, the probability that  $x_{12}$  of these channels makes a transition to the open unbound state and  $x_{14}$  channels make a transition to the closed bound state in the time interval  $[t, t + \Delta t)$  is

$$p(x_1, x_{12}, x_{14}) = M(x_1, k_{12}\Delta t, k_{14}\Delta t), \quad (56)$$

$$\approx B(x_1, k_{12}\Delta t)B(x_1, k_{14}\Delta t), \quad (57)$$

where  $M(n, p, p')$  is the multinomial distribution with success probabilities  $p$  and  $p'$  in  $n$  trials and  $B(n, p)$  is the binomial distribution with success probability  $p$  in  $n$  trials. We note that this method is equivalent to the binomial  $\tau$ -leaping method (78).

Computationally, we use Poisson or Gaussian approximations to the binomial distribution. Which approximation works better for  $B(n, p)$  depends both on  $n$  and  $p$ , and in our simulations both of these may change on each time step (in particular during sparks), so we determined before the simulations the region in the  $(n, p)$  space where each approximation was to be used. Denoting  $B_{n,p}(m) = (n/m)p^m(1-p)^{n-m}$ ,  $P_{n,p}(m) = (np)^m e^{-np}/m!$ , and  $G_{n,p}(m) = e^{-(m-np)^2/(2np(1-p))}/\sqrt{2\pi np(1-p)}$ , the probabilities of choosing  $m$  using the binomial, Poisson, and Gaussian distributions, we define for each  $n$  and  $p$  the errors

$$E^P(n, p) = \frac{1}{n} \sum_{m=0}^n (B_{n,p}(m) - P_{n,p}(m))^2, \quad (58)$$

$$E^G(n, p) = \frac{1}{n} \sum_{m=0}^n (B_{n,p}(m) - G_{n,p}(m))^2, \quad (59)$$

with the intent to use a Gaussian approximation when  $E^G(n, p) < E^P(n, p)$  and a Poisson approximation when  $E^G(n, p) \geq E^P(n, p)$ . To simplify the computations, we approximate the numerically found region in which  $E^P(n, p) < E^G(n, p)$  by the set of points  $(n, p)$  satisfying  $100p < 6.6 + 52n^{-1/2}$ , and use a Poisson approximation in this case and a Gaussian approximation otherwise. The Poisson distributed numbers are generated using the Knuth algorithm and the Gaussian distributed numbers with the Box-Mueller algorithm (79).

In summary, at each time step we have  $x_i$ , the number of RyR channels in state  $i$ . The number of channels  $x_{ij}$  making a transition from state  $i$  to state  $j$  is approximated by

$$x_{ij} = B(n_i, k_{ij}\Delta t), \quad (60)$$

where  $B$  is a random number generated to approximate a binomial distribution as described in the previous paragraph. After all the  $x_{ij}$  values have been

determined in this way, the number of RyRs in each state in the next step is updated as

$$x_i \rightarrow x_i - \sum_j x_{ij} + \sum_j x_{ji}, \quad (61)$$

where the sum is over the two states that are accessible to state  $i$ .

## Initial conditions and model integration

Initial conditions have to be supplied for the  $14N$  variables of our model. We find that, due to the intrinsic randomization caused by the stochastic nature of the model at the CRU level, the steady-state whole-cell behavior of the model during pacing does not depend on the initial conditions, although the first beats show transient behavior. In addition, even for identical initial conditions the simulation produces different results at single CRUs. Thus, the precise values of the initial conditions are not crucial when we are interested in steady-state behavior and we use the initial values  $c_i^{(n)} = c_s^{(n)} = c_p^{(n)} = 0.1 \mu\text{M}$ ,  $c_{\text{JSR}}^{(n)} = c_{\text{NSR}}^{(n)} = 750 \mu\text{M}$ ,  $x_1^{(n)} = 30$ ,  $x_2^{(n)} = 0$ ,  $x_3^{(n)} = 0$  (i.e., 70% of the RyR channels in the closed CSQN-bound state and 30% of the RyR channels in the closed CSQN-unbound state),  $y_j^{(n)} = 2$  (i.e., the L-type channels start in the closed two-state of Mahajan et al. (52)), and  $[\text{CaT}]_i^{(n)} = [\text{CaT}]_s^{(n)} = 20 \mu\text{M}$ . In any case, we always discard the first beats.

For the graded release curves (Fig. 3) and the spark restitution curves (Fig. 4) we start with the same initial conditions, except that  $c_{\text{JSR}}^{(n)} = c_{\text{NSR}}^{(n)} = 650 \mu\text{M}$ . In addition, we wait 1 s before collecting pairs of sparks in the restitution curve, and depolarizing to higher voltages in the graded release curve, to reach steady state and allow randomization to occur.

The SR depletion curve (Fig. 5) depends on how many RyR channels are in the CSQN-unbound state. We set the initial conditions to be the same as before, except that we set the initial SR and JSR content at  $400 \mu\text{M}$ . Then, we keep the myocyte at  $-80 \text{ mV}$  during various intervals and depolarize the cell, obtaining points with various values of SR content before release.

The fastest timescales in the model are diffusion from the proximal space to the submembrane space, and from the submembrane space to the cytosol. In the numerical integration of the model, it was assumed that equilibration of these processes is instantaneous (Eqs. 51 and 54). The fastest of the remaining timescales is the mean open time of the L-type channels,  $\sim 1 \text{ ms}$ , and we found that an adequate time step was  $\Delta t = 0.1 \text{ ms}$ . The results do not change if we use  $\Delta t = 0.02 \text{ ms}$ . The ODEs were integrated using the explicit Euler method and the Markov models were updated every timestep as described in the previous section.

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